CMSC 471 Artificial Intelligence

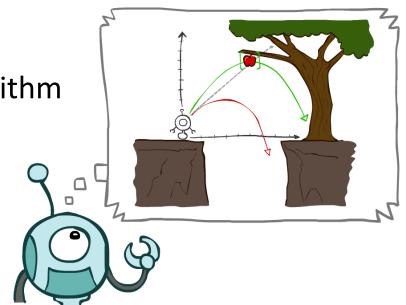
Search

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Some material adopted from notes by Charles R. Dyer, University of Wisconsin-Madison

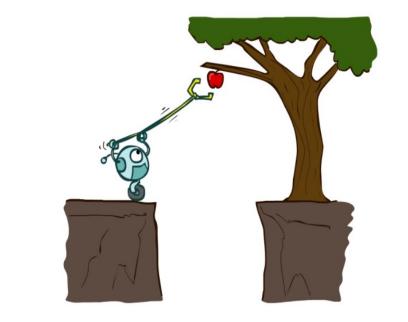
Today

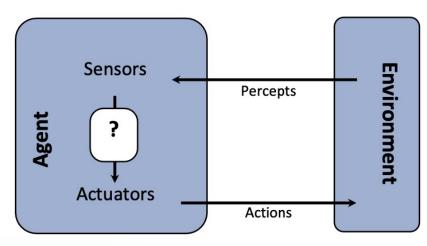
- Agents that Plan Ahead
- Goal-based agents
- Search Problems
- Generic state-space search algorithm
- Uninformed Search Methods
 - Depth-First Search
 - Breadth-First Search
 - Uniform-Cost Search



How do you design an intelligent agent?

- An agent is an entity that perceives and acts
- Intelligent agents perceive environment via sensors and act rationally on them with their effectors
- Discrete agents receive
 percepts one at a time, and
 map them to a sequence of
 discrete actions





Characteristics of environments

	Fully observable?	Deterministic?	Episodic?	Static?	Discrete?	Single agent?
Solitaire	No	Yes	Yes	Yes	Yes	Yes
Backgammon	Yes	No	No	Yes	Yes	No
Taxi driving	No	No	No	No	No	No
Internet shopping	No	No	No	No	Yes	No
Medical diagnosis	No	No	No	No	No	Yes

A Yes in a cell means that aspect is simpler; a No more complex

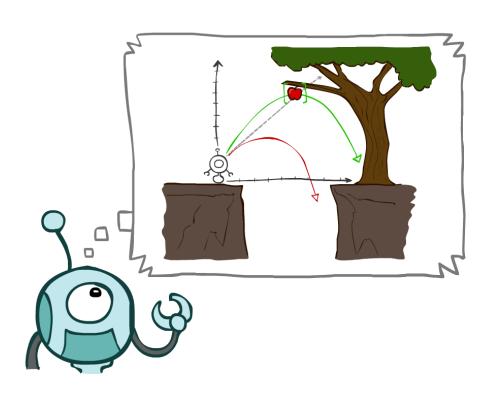
Characteristics of environments

→ Lots of real-world domains fall into the hardest case!

	Fully observable?	Deterministic?	Episodic?	Static?	Discrete?	Single agent?
Solitaire	No	Yes	Yes	Yes	Yes	Yes
Backgammon	Yes	No	No	Yes	Yes	No
Taxi driving	No	No	No	No	No	No
Internet shopping	No	No	No	No	Yes	No
Medical diagnosis	No	No	No	No	No	Yes

A Yes in a cell means that aspect is simpler; a No more complex

Agents that Plan



(0) Table-driven agents

simple

Use percept sequence/action table to find next action. Implemented by a **lookup table**

(1) Simple reflex agents

Based on **condition-action rules**, stateless devices with no memory of past world states

(2) Agents with memory

represent states and keep track of past world states

(3) Agents with goals

Have a state and **goal information** describing desirable situations; can take future events into consideration

(4) Utility-based agents

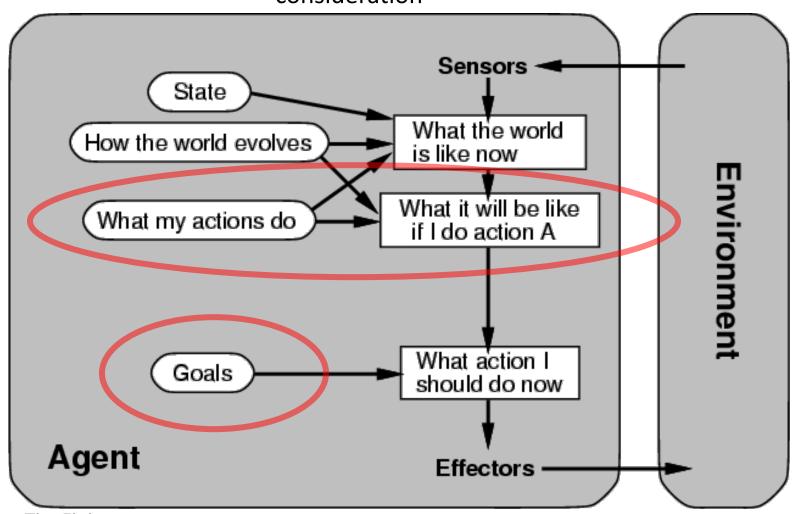
base decisions on utility theory in order to act rationally





(3) Architecture for goal-based agent

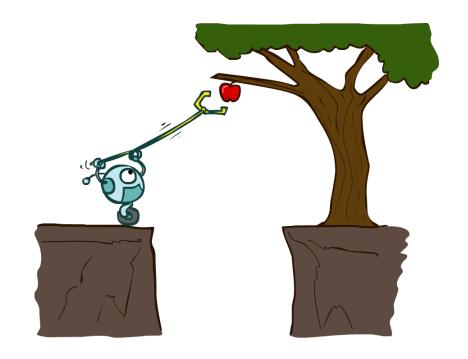
state and **goal information** describe desirable situations allowing agent to take future events into consideration



Planning Agents

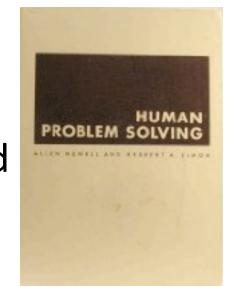
Planning agents:

- Ask "what if"
- Decisions based on (hypothesized) consequences of actions
- Must have a model of how the world evolves in response to actions
- Must formulate a goal (test)
- Consider how the world WOULD BE
- Optimal vs. complete planning



Big Idea

Allen Newell and Herb Simon developed the *problem space principle* as an Al approach in the late 60s/early 70s

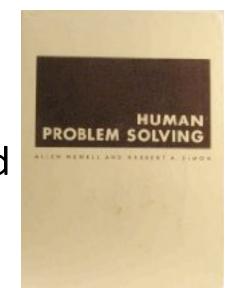


"The rational activity in which people engage to solve a problem can be described in terms of (1) a set of **states** of knowledge, (2) **operators** for changing one state into another, (3) **constraints** on applying operators and (4) **control** knowledge for deciding which operator to apply next."

Newell A & Simon H A. Human problem solving. Englewood Cliffs, NJ: Prentice-Hall. 1972.

Big Idea

Allen Newell and Herb Simon developed the *problem space principle* as an Al approach in the late 60s/early 70s



nto

"The rat problem of know another control next."

We'll achieve this by formulating an appropriate graph and then applying graph search algorithms to it

Newell A & Simon H A. Human problem solving. Englewood Cliffs, NJ: Prentice-Hall. 1972.

A search problem consists of:

- A search problem consists of:
 - A state space













- A search problem consists of:
 - A state space















A successor function (with actions, costs)

- A search problem consists of:
 - A state space







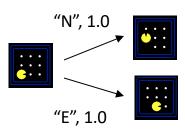








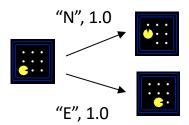
A successor function (with actions, costs)



- A search problem consists of:
 - A state space



A successor function (with actions, costs)

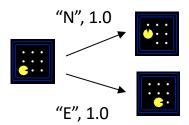


A start state and a goal test

- A search problem consists of:
 - A state space



A successor function (with actions, costs)



- A start state and a goal test
- A solution is a sequence of actions (a plan) which transforms the start state to a goal state

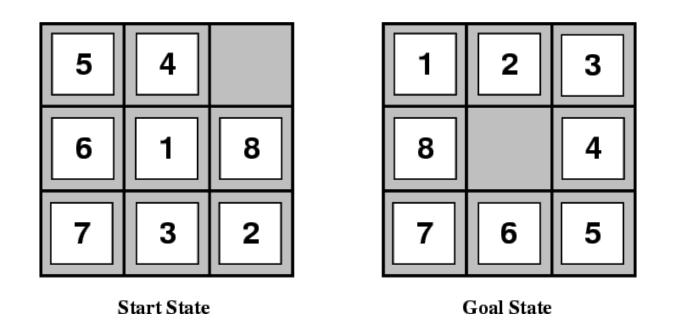
Search Problems Are Models

Search Problems Are Models

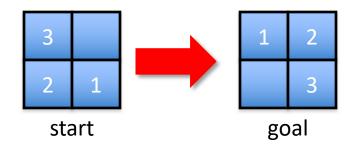


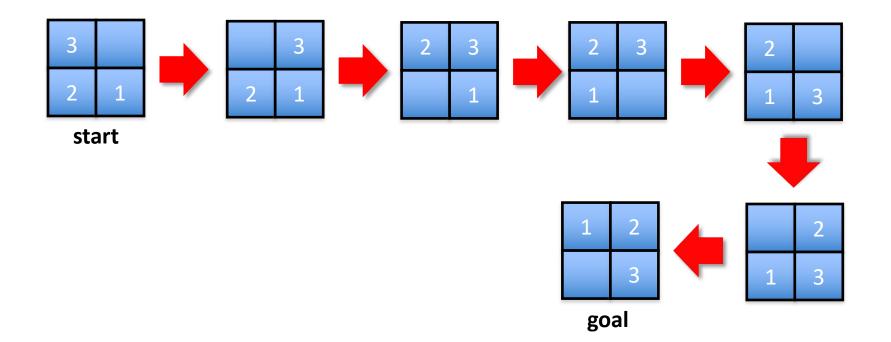
Example: 8-Puzzle

Given an initial configuration of 8 numbered tiles on a 3x3 board, move the tiles to produce a desired goal configuration



Simpler: 3-Puzzle

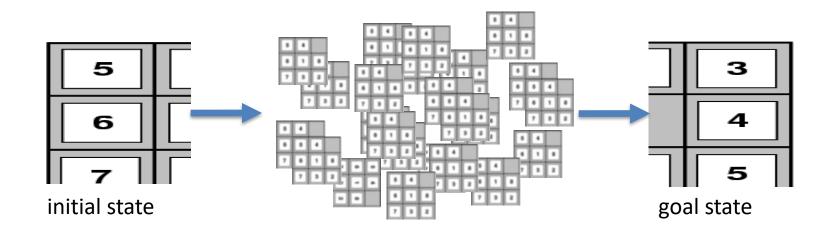




Building goal-based agents

We must answer the following questions

- -How do we represent the state of the "world"?
- -What is the **goal** and how can we recognize it?
- –What are the possible actions?
- –What relevant information do we encoded to describe states, actions and their effects and thereby solve the problem?



Characteristics of 8-puzzle?

	Fully observable?	Deterministic?	Episodic?	Static?	Discrete?	Single agent?
8-puzzle						

Characteristics of 8-puzzle

	Fully observable?	Deterministic?	Episodic?	Static?	Discrete?	Single agent?
8-puzzle	Yes	Yes	Yes	Yes	Yes	Yes

- All the Yes's mean it may be relatively easy!
- This is typical of the problems worked on in the 60s and 70s
- And the algorithms for solving them a statespace search model

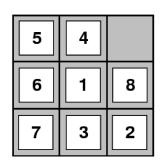
Representing states / State-space

• State of an 8-puzzle?

5	4	
6	1	8
7	3	2

Representing states / State-space

- State of an 8-puzzle?
 - A 3x3 array of integer in {0..8}
 - No integer appears twice
 - 0 represents the empty space



- In Python, we might implement this using a nine-character string: "540681732"
- And write functions to map the 2D coordinates to an index

What's the goal to be achieved?



- Describe situation we want to achieve, a set of properties that we want to hold, etc.
- Defining a goal test function that when applied to a state returns True or False
- For our problem:

```
def isGoal(state):
    return state == "123405678"
```

What are the actions?



- Primitive actions for changing the state
 In a deterministic world: no uncertainty in an action's effects (simple model)
- Given action and description of current world state, action completely specifies
 - Whether action can be applied to the current world (i.e., is it applicable and legal?) and
 - What state *results* after action is performed in the current world (i.e., no need for *history* information to compute the next state)

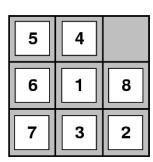
Representing actions



- Actions ideally considered as discrete events that occur at an instant of time
- Example, in a planning context
 - If state:inClass and perform action:goHome, then next state is state:atHome
 - There's no time where you're neither in class nor at home (i.e., in the state of "going home")

Representing actions

Actions for 8-puzzle?



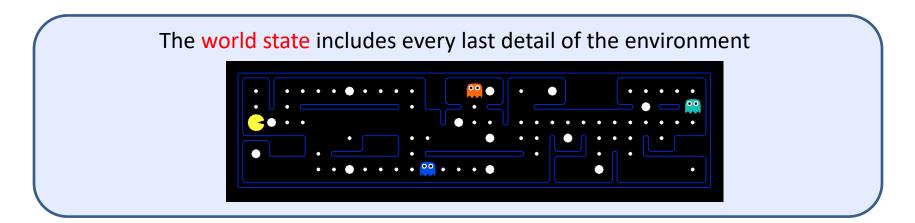
Representing actions

 5
 4

 6
 1
 8

 7
 3
 2

- Actions for 8-puzzle?
- Number of actions/operators depends on the representation used in describing a state
 - Specify 4 possible moves for each of the 8 tiles,
 resulting in a total of 4*8=32 operators
 - Or: Specify four moves for "blank" square and we only need 4 operators
- Representational shift can simplify a problem!



The world state includes every last detail of the environment



A search state keeps only the details needed for planning (abstraction)

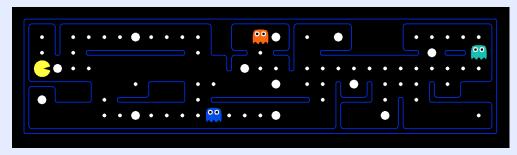
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A search state keeps only the details needed for planning (abstraction)

Problem: Pathing

The world state includes every last detail of the environment



A search state keeps only the details needed for planning (abstraction)

- Problem: Pathing
 - States: (x,y) location

The world state includes every last detail of the environment



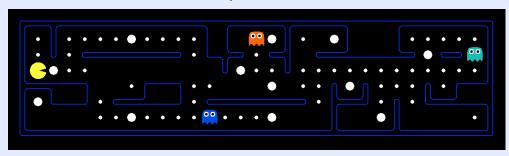
A search state keeps only the details needed for planning (abstraction)

Problem: Pathing

States: (x,y) location

Actions: NSEW

The world state includes every last detail of the environment



A search state keeps only the details needed for planning (abstraction)

Problem: Pathing

States: (x,y) location

Actions: NSEW

Successor: update location

only

The world state includes every last detail of the environment



A search state keeps only the details needed for planning (abstraction)

Problem: Pathing

States: (x,y) location

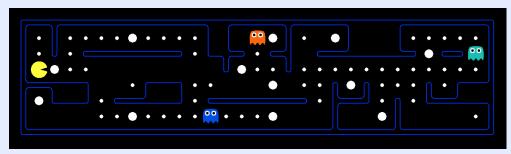
Actions: NSEW

Successor: update location

only

Goal test: is (x,y)=END

The world state includes every last detail of the environment



A search state keeps only the details needed for planning (abstraction)

Problem: Pathing

States: (x,y) location

Actions: NSEW

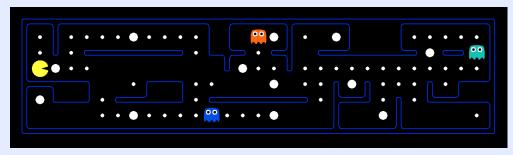
Successor: update location

only

Goal test: is (x,y)=END

Problem: Eat-All-Dots

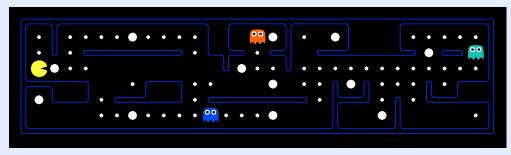
The world state includes every last detail of the environment



- Problem: Pathing
 - States: (x,y) location
 - Actions: NSEW
 - Successor: update location only
 - Goal test: is (x,y)=END

- Problem: Eat-All-Dots
 - States: {(x,y), dot booleans}

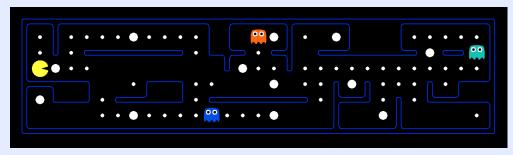
The world state includes every last detail of the environment



- Problem: Pathing
 - States: (x,y) location
 - Actions: NSEW
 - Successor: update location only
 - Goal test: is (x,y)=END

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 - States: {(x,y), dot booleans}
 - Actions: NSEW

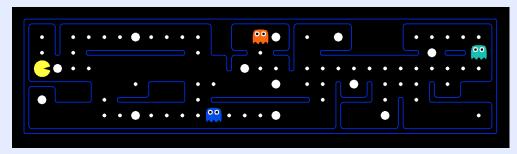
The world state includes every last detail of the environment



- Problem: Pathing
 - States: (x,y) location
 - Actions: NSEW
 - Successor: update location only
 - Goal test: is (x,y)=END

- Problem: Eat-All-Dots
 - States: {(x,y), dot booleans}
 - Actions: NSEW
 - Successor: update location and possibly a dot boolean

The world state includes every last detail of the environment



- Problem: Pathing
 - States: (x,y) location
 - Actions: NSEW
 - Successor: update location only
 - Goal test: is (x,y)=END

- Problem: Eat-All-Dots
 - States: {(x,y), dot booleans}
 - Actions: NSEW
 - Successor: update location and possibly a dot boolean
 - Goal test: dots all false

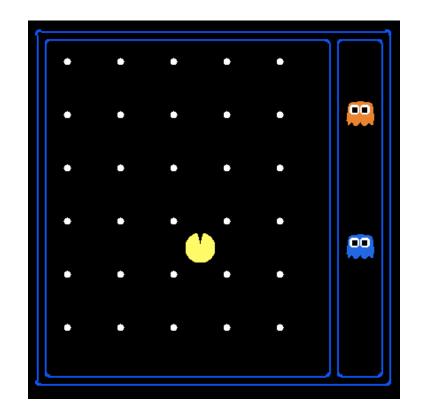
World state:

Agent positions: 120

Food count: 30

Ghost positions: 12

Agent facing: NSEW



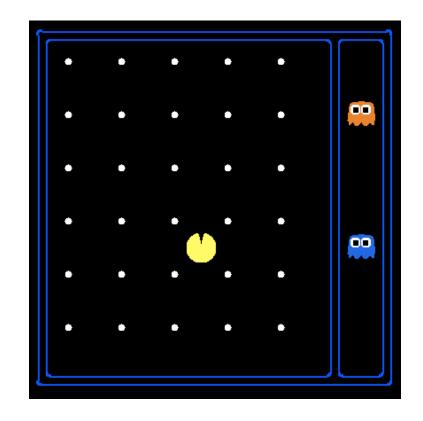
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World state:

Agent positions: 120

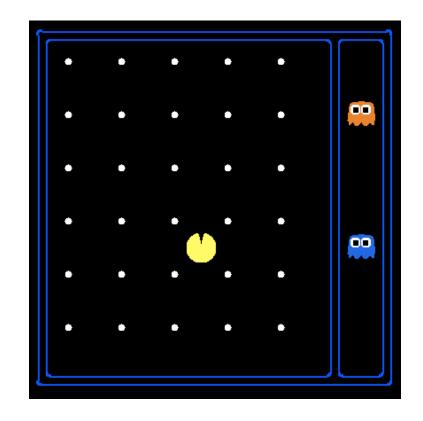
Food count: 30

Ghost positions: 12

Agent facing: NSEW

How many

– World states?



World state:

Agent positions: 120

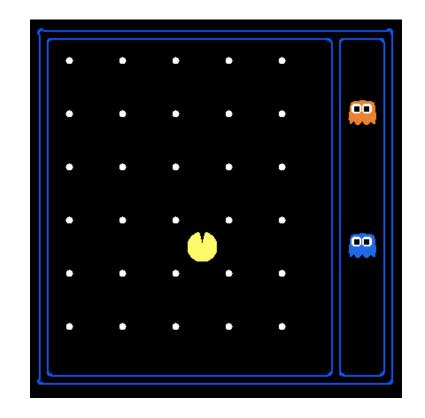
Food count: 30

Ghost positions: 12

Agent facing: NSEW

How many

World states?
 120x(2³⁰)x(12²)x4



World state:

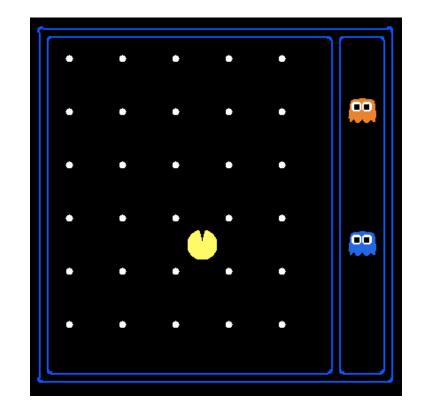
Agent positions: 120

Food count: 30

Ghost positions: 12

Agent facing: NSEW

- World states?120x(2³⁰)x(12²)x4
- States for pathing?



World state:

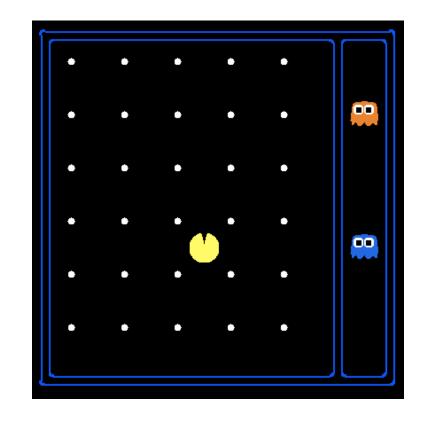
Agent positions: 120

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Agent facing: NSEW

- World states?
 120x(2³⁰)x(12²)x4
- States for pathing?120



World state:

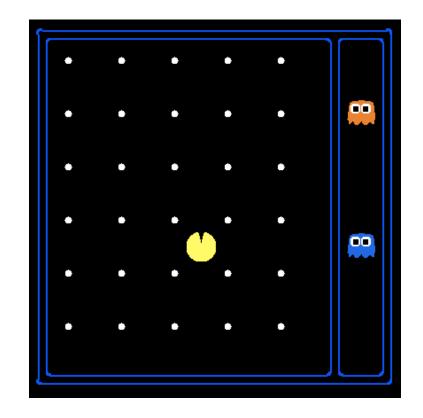
Agent positions: 120

Food count: 30

Ghost positions: 12

Agent facing: NSEW

- World states?
 120x(2³⁰)x(12²)x4
- States for pathing?120
- States for eat-all-dots?



World state:

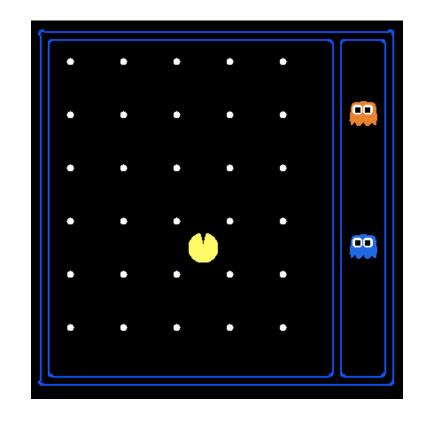
Agent positions: 120

Food count: 30

Ghost positions: 12

Agent facing: NSEW

- World states?
 120x(2³⁰)x(12²)x4
- States for pathing?120
- States for eat-all-dots?120x(2³⁰)





- Size of a problem usually described in terms of possible number of states
 - Tic-Tac-Toe has about 3⁹ states (19,683≈2*10⁴)
 - Checkers has about 10⁴⁰ states
 - Rubik's Cube has about 10¹⁹ states
 - Chess has about 10¹²⁰ states in a typical game
 - Go has $2*10^{170}$
 - Theorem provers may deal with an infinite space
- State space size ≈ solution difficulty



- Our estimates were loose upper bounds
- How many possible, legal states does tictac-toe really have?
- Simple upper bound: nine board cells, each of which can be empty, O or X, so 39
- Only 593 states after eliminating
 - impossible states
 - Rotations and reflections

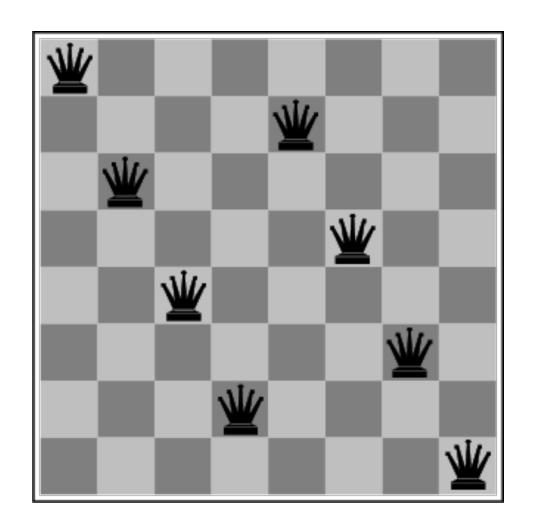
Some example problems

- Toy problems and micro-worlds
 - -8-Puzzle
 - Missionaries and Cannibals
 - Cryptarithmetic
 - -8-Queens Puzzle
 - Remove 5 Sticks
 - Water Jug Problem
- Real-world problems

Example: The 8-Queens Puzzle

Place eight queens on a chessboard such that no queen attacks any other

We can generalize the problem to a NxN chessboard

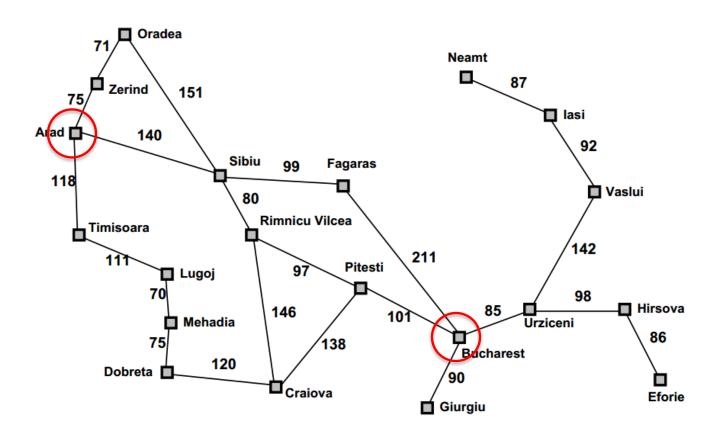


Some more real-world problems

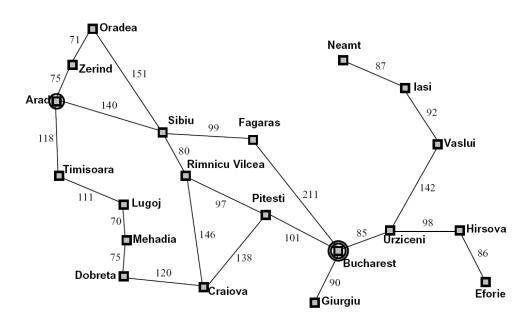
- Route finding
- Touring (traveling salesman)
- Logistics
- VLSI layout
- Robot navigation
- Theorem proving
- Learning

Route Planning

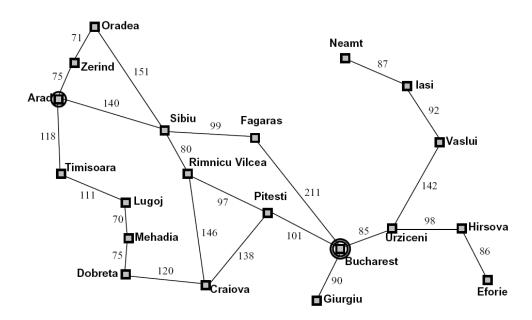
Find a route from Arad to Bucharest



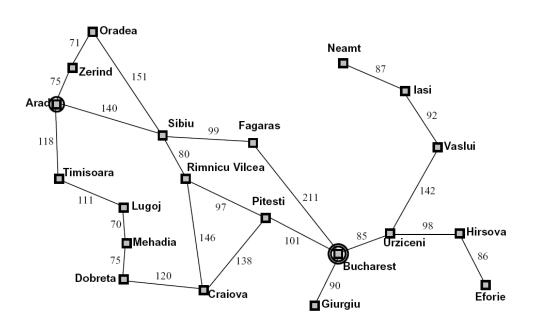
A simplified map of major roads in Romania used in our text

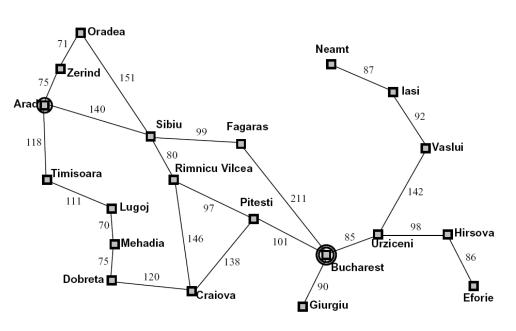


• State space:

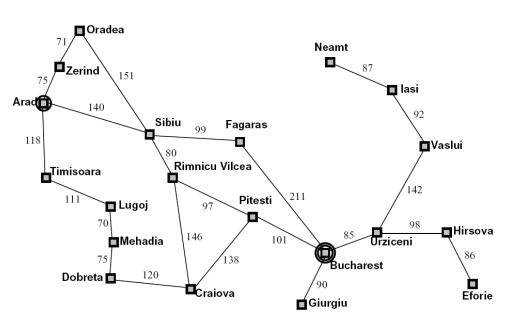


- State space:
 - Cities

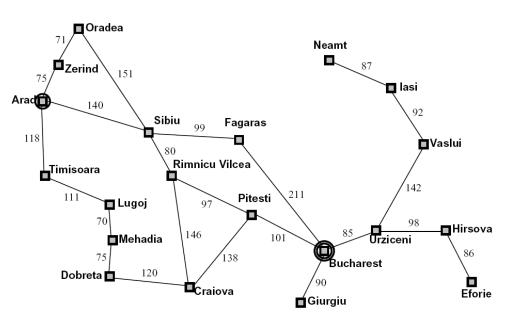




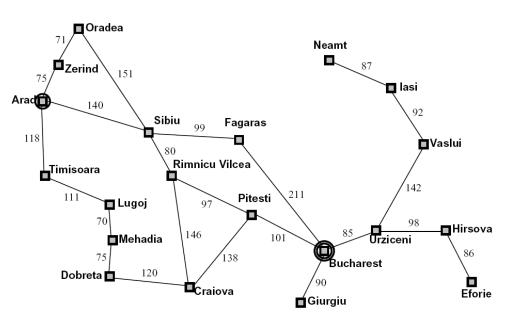
- State space:
 - Cities
- Successor function:



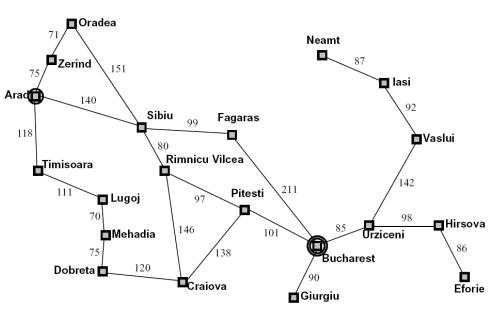
- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance



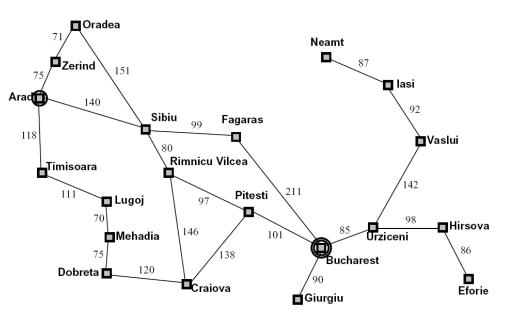
- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance
- Start state:



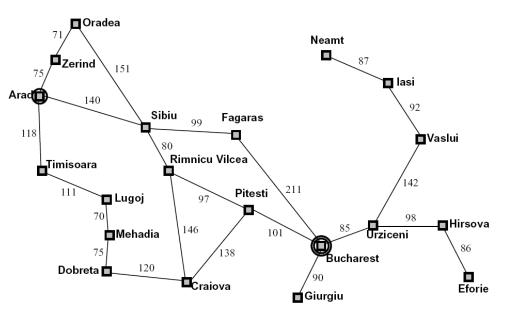
- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance
- Start state:
 - Arad



- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance
- Start state:
 - Arad
- Goal test:

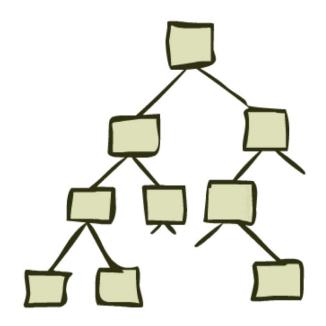


- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance
- Start state:
 - Arad
- Goal test:
 - Is state == Bucharest?



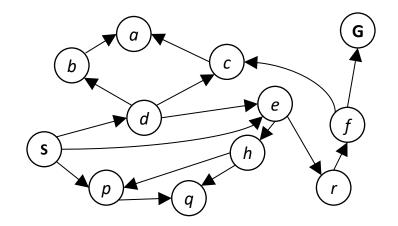
- State space:
 - Cities
- Successor function:
 - Roads: Go to adjacent citywith cost = distance
- Start state:
 - Arad
- Goal test:
 - Is state == Bucharest?
- Solution?

State Space Graphs and Search Trees



State Space Graphs

- State space graph: A mathematical representation of a search problem
 - Nodes are (abstracted) world configurations
 - Arcs represent transitions/ successors (action results)
 - The goal test is a set of goal nodes (maybe only one)
- In a state space graph, each state occurs only once!
- We can rarely build this full graph in memory (it's too big), but it's a useful idea



Tiny state space graph for a tiny search problem

Search Trees

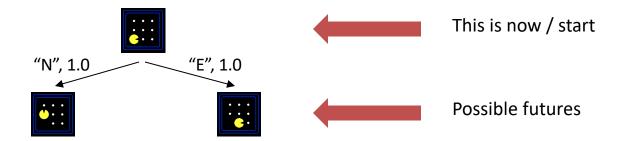
Search Trees



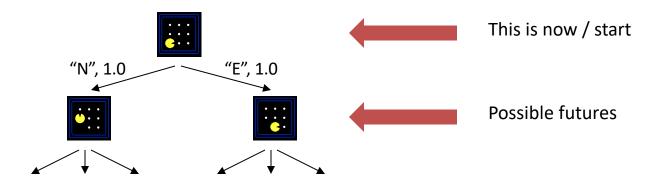


This is now / start

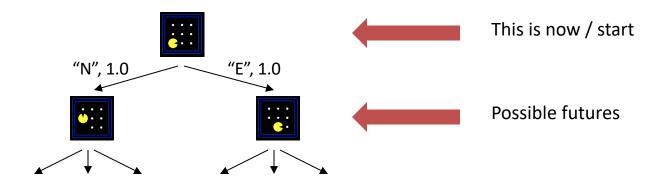
Search Trees



Search Trees

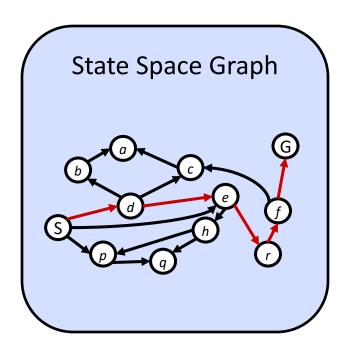


Search Trees



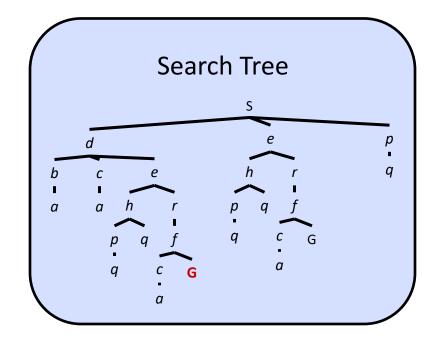
A search tree:

- A "what if" tree of plans and their outcomes
- The start state is the root node
- Children correspond to successors
- Nodes show states, but correspond to PLANS that achieve those states
- For most problems, we can never actually build the whole tree

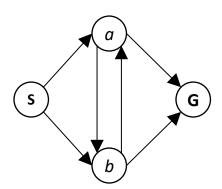


Each NODE in in the search tree is an entire PATH in the state space graph.

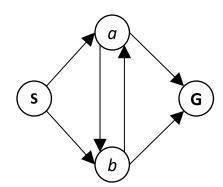
We construct the tree on demand — and we construct as little as possible.



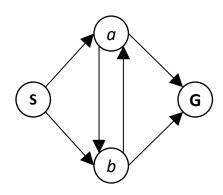
Consider this 4-state graph:



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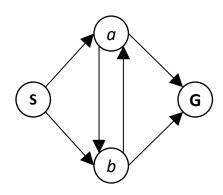


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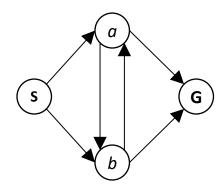
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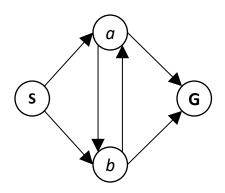
How big is its search tree (from S)?



S



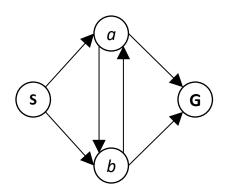
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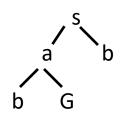






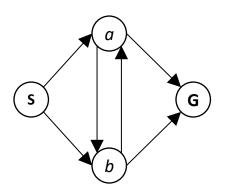
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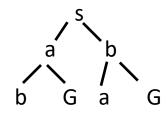






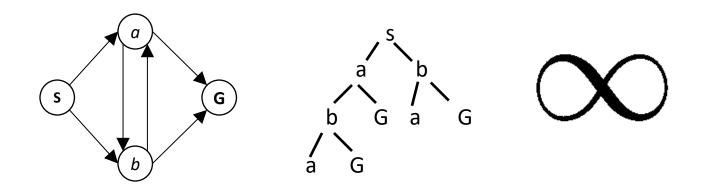
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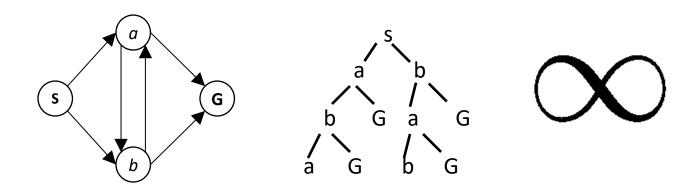




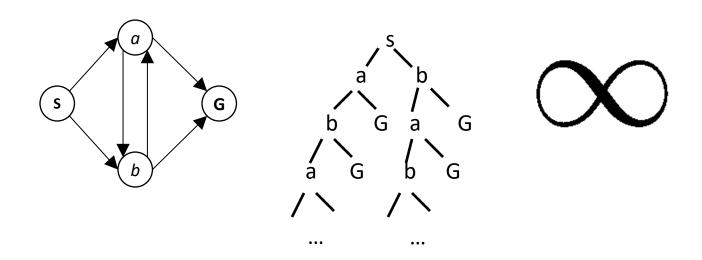
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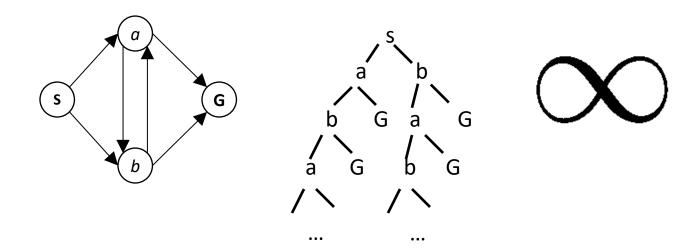


Consider this 4-state graph:



Consider this 4-state graph:

How big is its search tree (from S)?



Important: Those who don't know history are doomed to repeat it!

Formalizing search

- Solution: sequence of actions associated with a path from a start node to a goal node
- Solution cost: sum of the arc costs on the solution path
 - If all arcs have same (unit) cost, then solution cost is length of solution (number of steps)
 - Algorithms generally require that arc costs cannot be negative (why?)

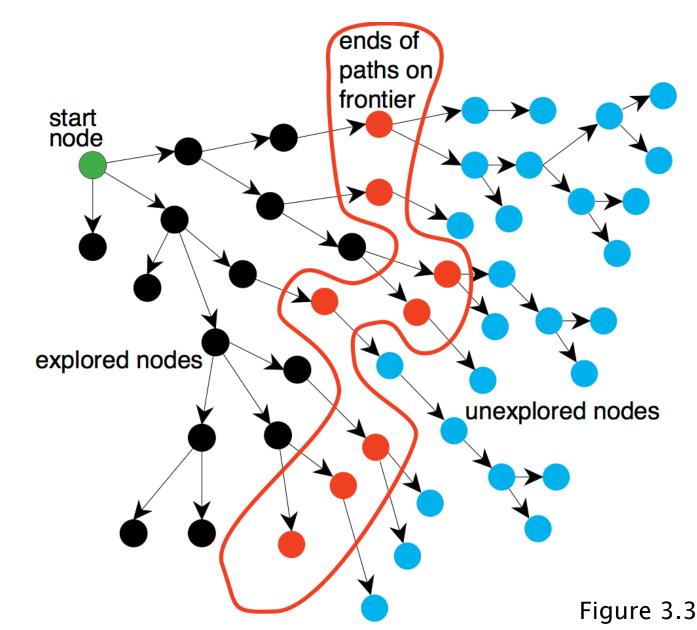
Formalizing search

- State-space search: searching through state space for solution by making explicit a portion of an implicit state-space graph to find a goal node
 - Can't materialize whole space for large problems
 - Initially V={S}, where S is the start node, E={}
 - On expanding S, its successor nodes are generated and added to V and associated arcs added to E
 - Process continues until a goal node is found
- Nodes represent a partial solution path (+ cost of partial solution path) from S to the node
 - From a node there may be many possible paths (and thus solutions) with this partial path as a prefix

A General Searching Algorithm

Core ideas:

- Maintain a list of frontier (fringe) nodes
 - Nodes coming into the frontier have been explored
 - 2. Nodes going out of the frontier have not been explored
- 2. Iteratively select nodes from the frontier and explore unexplored nodes from the frontier
- 3. Stop when you reach your **goal**



State-space search algorithm

```
;; problem describes the start state, operators, goal test, and operator costs
;; queueing-function is a comparator function that ranks two states
;; general-search returns either a goal node or failure
function general-search (problem, QUEUEING-FUNCTION)
  nodes = MAKE-QUEUE (MAKE-NODE (problem.INITIAL-STATE) )
  loop
      if EMPTY(nodes) then return "failure"
      node = REMOVE-FRONT(nodes)
      if problem.GOAL-TEST(node.STATE) succeeds
          then return node
      nodes = QUEUEING-FUNCTION(nodes, EXPAND(node,
                problem.OPERATORS))
 end
  ;; Note: The goal test is NOT done when nodes are generated
  ;; Note: This algorithm does not detect loops
```

Key procedures to be defined

EXPAND

 Generate a node's successor nodes, adding them to the graph if not already there

GOAL-TEST

Test if state satisfies all goal conditions

QUEUEING-FUNCTION

- Maintain ranked list of nodes that are candidates for expansion
- Changing definition of the QUEUEING-FUNCTION leads to different search strategies

What does "search" look like for a particular problem?

start

1	2	3
4	8	
7	6	5

goal



start

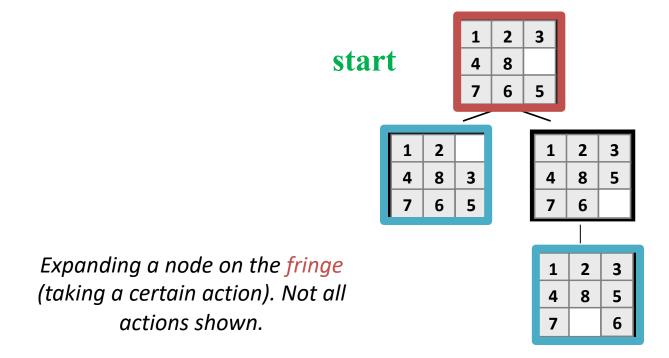
2 8 6

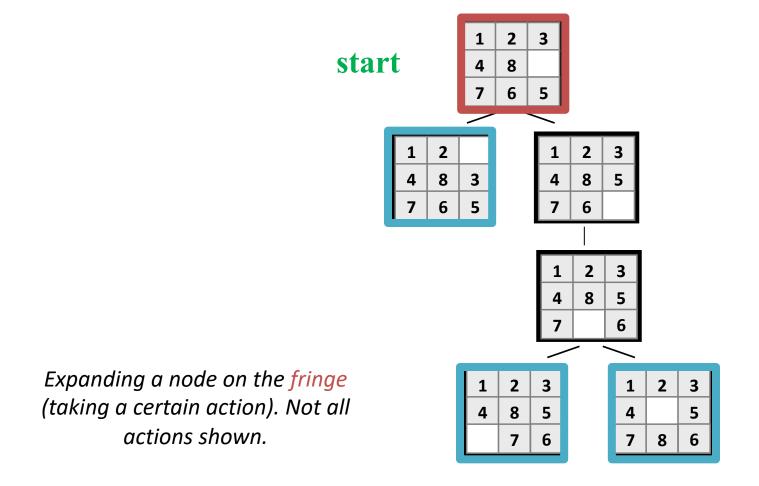
Expanding a node on the fringe (taking a certain action)

1	2	
4	8	3
7	6	5

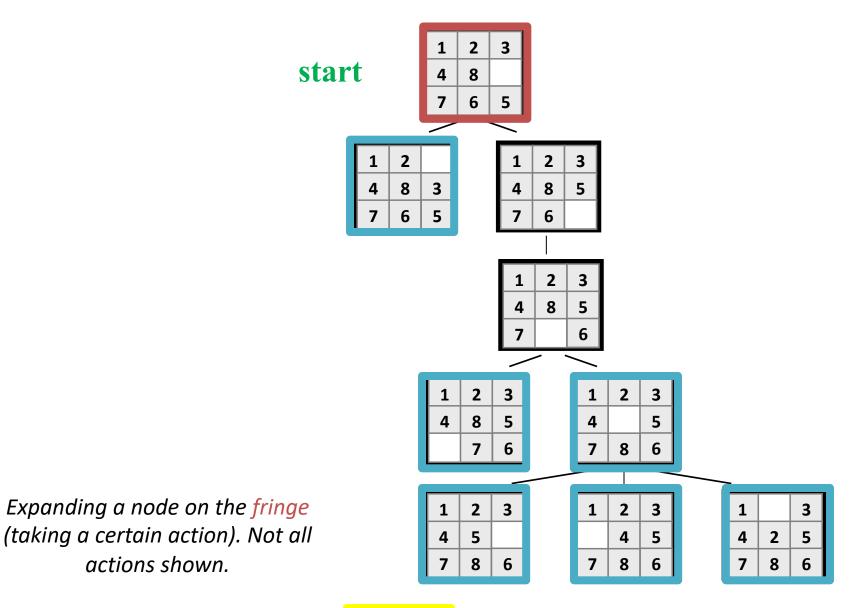
1	2	3
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7	6	



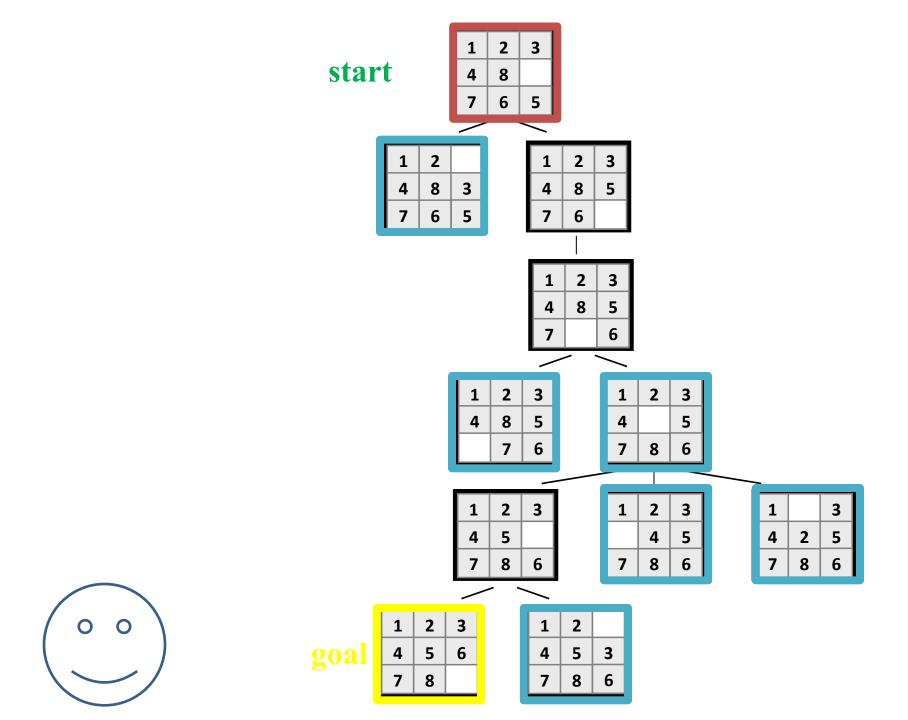








1 2 3 4 5 6 7 8



Informed vs. uninformed search



Uninformed search strategies (blind search)

- -Use no information about likely direction of a goal
- Methods: breadth-first, depth-first, depth-limited, uniform-cost, depth-first iterative deepening, bidirectional

Informed search strategies (heuristic search)

- Use information about domain to (try to) (usually)
 head in the general direction of goal node(s)
- Methods: hill climbing, best-first, greedy search,
 beam search, algorithm A, algorithm A*

Evaluating search strategies

- Completeness
 - Guarantees finding a solution whenever one exists
- Time complexity (worst or average case)
- Space complexity
- Optimality/Admissibility

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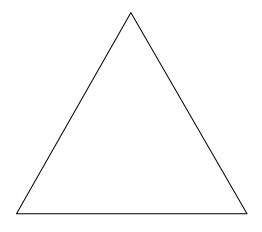
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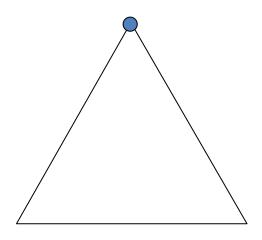
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- Cartoon of search tree:

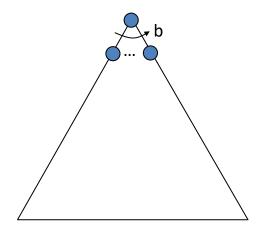
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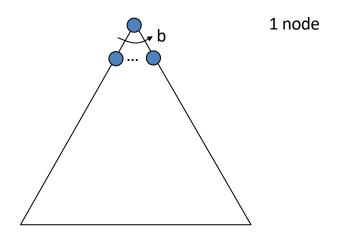
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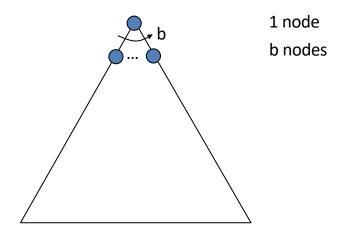
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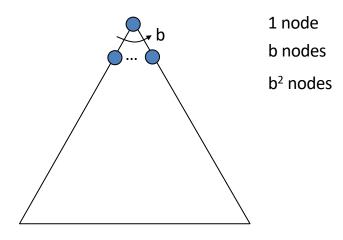
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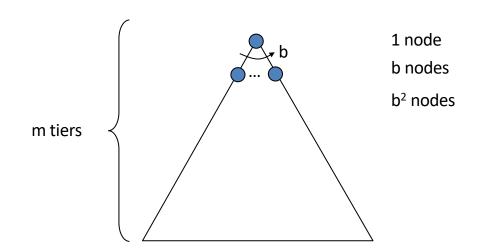
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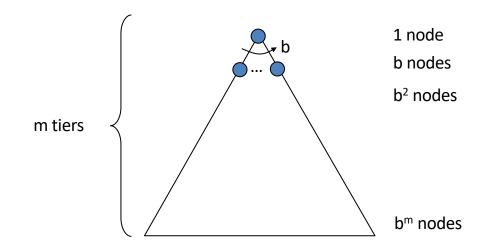
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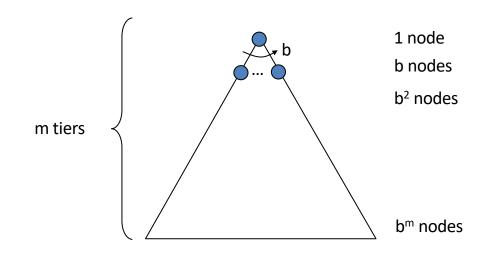
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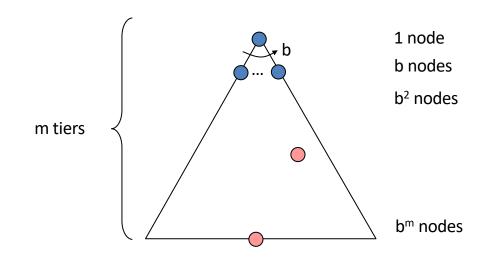
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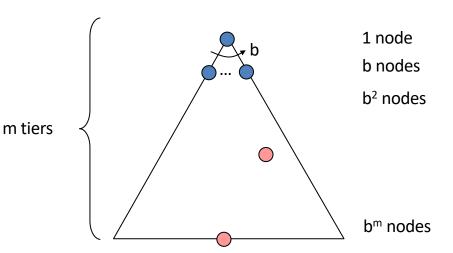
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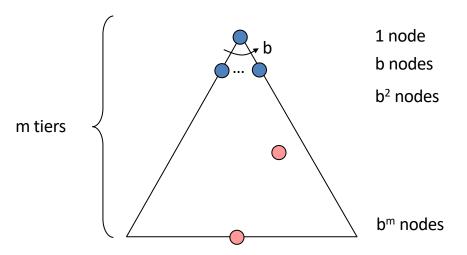
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 - 1 + b + b² + b^m = O(b^m)



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Optimality/Admissibility

 If a solution is found, is it guaranteed to be an optimal one, i.e., one with minimum cost



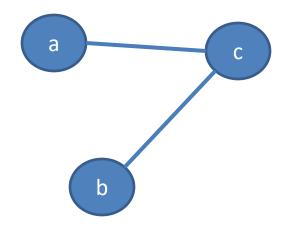
- A graph G = (E, V)
- V = set of vertices (nodes)
- E = set of edges between pairs of nodes, (x, y)

G can be:

- Undirected: order of (x, y) doesn't matter
 - These are symmetric
- Directed: order of (x, y) does matter
- Weighted: cost function g(x, y)
- (among other qualities)

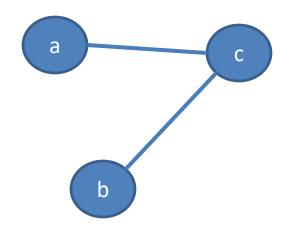


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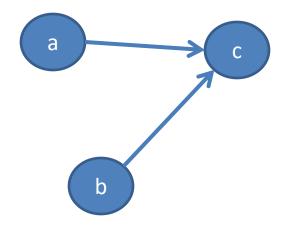
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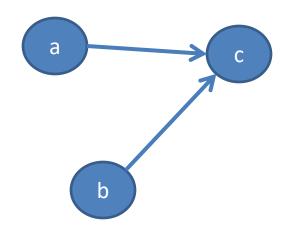


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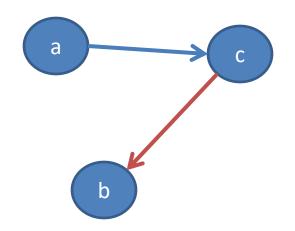
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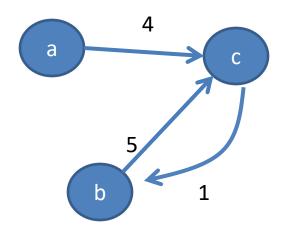
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$$E = \{ (a, c), (c, b) \}$$



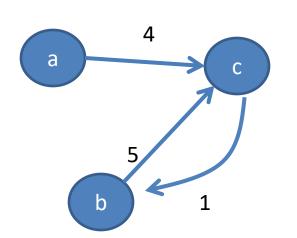
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$$g = ???$$



- A graph G = (E, V)
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$$E = \{ (a,c), (b, c), (c, b) \}$$

$$g = \{(a, c): 4, (b, c): 5, (c, b): 1\}$$

Water Jug Problem

- Two jugs J1 & J2 with capacity C1 & C2
- Initially J1 has W1 water and J2 has W2 water
 - e.g.: full 5 gallon jug and empty 2 gallon jug
- Possible actions:
 - Pour from jug X to jug Y until X empty or Y full
 - Empty jug X onto the floor
- Goal: J1 has G1 water and J2 G2
 - G1 or G2 can be -1 to represent any amount
- E.g.: initially full jugs with capacities 3 and 1 liters, goal is to have 1 liter in each



Example: Water Jug Problem



- Two jugs J1 and J2 with capacity C1 and C2
- Initially J1 has W1 water and J2 has W2 water
 - e.g.: a full 5-gallon jug and an empty 2-gallon jug
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Example: Water Jug Problem



Given full 5-gal. jug and empty 2-gal. jug, fill 2-gal jug with one gallon

- State representation?
 - -General state?
 - —Initial state?
 - -Goal state?
- Possible actions?
 - -Condition?
 - -Resulting state?

Action table

Name	Cond.	Transition	Effect

Example: Water Jug Problem



Given full 5-gal. jug and empty 2-gal. jug, fill 2-gal jug with one gallon

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Action table

Name	Cond.	Transition	Effect
dump1	x>0	$(x,y) \rightarrow (0,y)$	Empty Jug 1
dump2	y>0		Empty Jug 2
pour_1_2	x>0 & y <c2< td=""><td></td><td>Pour from Jug 1 to Jug 2</td></c2<>		Pour from Jug 1 to Jug 2
pour_2_1	y>0 & X <c1< td=""><td></td><td>Pour from Jug 2 to Jug 1</td></c1<>		Pour from Jug 2 to Jug 1

So...

- How can we represent the states?
- What's an initial state
- How do we recognize a goal state
- What are the actions; how can we tell which ones can be performed in a given state; what is the resulting state
- How do we search for a solution from an initial state given a goal state
- What is a solution? The goal state achieved or a path to it?

Search in a state space

- Basic idea:
 - Create representation of initial state
 - -Try all possible actions & connect states that result
 - Recursively apply process to the new states until we find a solution or dead ends
- We need to keep track of the connections between states and might use a
 - —Tree data structure or
 - Graph data structure
- A graph structure is best in general...

Formalizing state space search

- A state space is a graph (V, E) where V is a set of nodes and E is a set of arcs, and each arc is directed from a node to another node
- **Nodes:** data structures with state description and other info, e.g., node's parent, name of action that generated it from parent, etc.
- Arcs: instances of actions, head is a state, tail
 is the state that results from action

Formalizing search in a state space

- Each arc has fixed, positive cost associated with it corresponding to the action cost
 - Simple case: all costs are 1
- Each node has a set of successor nodes corresponding to all legal actions that can be applied at node's state
 - Expanding a node = generating its successor nodes and adding them and their associated arcs to the graph
- One or more nodes are marked as start nodes
- A goal test predicate is applied to a state to determine if its associated node is a goal node