

Solving Problems by Searching

AIMA chapter 2 (partly), AIMA Sections 3.1–3.3

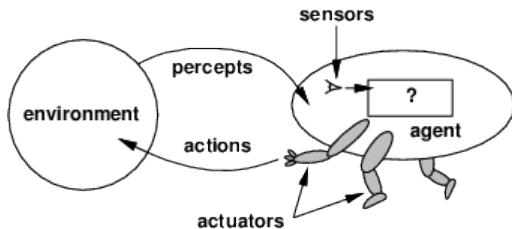
Outline

Solving
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- ◇ Rational agents
- ◇ Problem-solving agents
- ◇ Problem types
- ◇ Problem formulation
- ◇ Example problems
- ◇ General search algorithm

Agents and environments

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Agents include humans, robots, softbots, thermostats, etc.

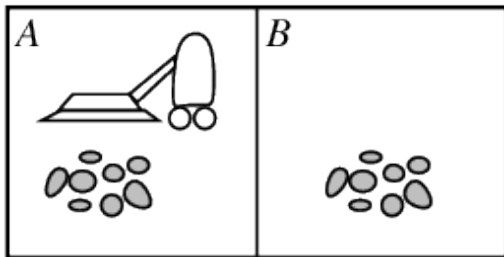
The **agent function** maps from percept histories to actions:

$$f : \mathcal{P}^* \rightarrow \mathcal{A}$$

The **agent program** runs on the physical **architecture** to produce f

Example: Vacuum-cleaner world

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Perceptions: location and contents, e.g., $[A, \textit{Dirty}]$

Actions: *Left*, *Right*, *Suck*, *NoOp*

A vacuum-cleaner agent

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Percept sequence	Action
<i>[A, Clean]</i>	<i>Right</i>
<i>[A, Dirty]</i>	<i>Suck</i>
<i>[B, Clean]</i>	<i>Left</i>
<i>[B, Dirty]</i>	<i>Suck</i>
<i>[A, Clean], [A, Clean]</i>	<i>Right</i>
<i>[A, Clean], [A, Dirty]</i>	<i>Suck</i>
<i>⋮</i>	<i>⋮</i>

What is the **right** function?

Can it be implemented in a **small** agent program?

Agent Programs vs. Agent Functions

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Question

If an agent has $|\mathcal{P}|$ possible perceptions, how many entries will the agent function have after T time steps ?

Agent Programs vs. Agent Functions

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Sol

$$\sum_{t=1}^T |\mathcal{P}|^t$$

AI goal \Rightarrow Design **small** agent programs to represent huge agent functions

A possible agent program

```
function Reflex-Vacuum-Agent( [location,status]) returns an action
```

```
  if status = Dirty then return Suck  
  else if location = A then return Right  
  else if location = B then return Left
```


Rationality

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Fixed **performance measure** evaluates the **environment sequence**

- one point per square cleaned up in time T ?
- one point per clean square per time step, minus one per move?
- penalize for $> k$ dirty squares?

A **rational agent** chooses whichever action maximizes the **expected** value of the performance measure **given the percept sequence to date**

Rational \neq omniscient

- percepts may not supply all relevant information

Rational \neq clairvoyant

- action outcomes may not be as expected

Hence, rational \neq successful

Rational \implies exploration, learning, autonomy

Multi-Robot Patrolling

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Exercise

Consider the following environment:

- Three rooms (A,B,C) and two robots (r_1, r_2)
- r_1 can patrol A and B, r_2 can patrol B and C
- r_1 starts from A and r_2 starts from C
- travel time between rooms is zero
- Performance measure: minimise sum of all rooms' average Idleness
- Average idleness = sum of time interval for which the room is not visited by any robot / total time interval

What would be a rational behavior for this environment ?

To design a rational agent, we must specify the **task environment**

Consider, e.g., the task of designing an automated taxi:

Performance measure??

Environment??

Actuators??

Sensors??

To design a rational agent, we must specify the **task environment**

Consider, e.g., the task of designing an automated taxi:

Performance measure?? safety, destination, profits, legality, comfort, ...

Environment?? city streets/freeways, traffic, pedestrians, weather, ...

Actuators?? steering, accelerator, brake, horn, speaker/display, ...

Sensors?? video, accelerometers, gauges, engine sensors, keyboard, GPS, ...

Environment types

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	Crossword	Robo-selector	Poker	Taxi
<u>Observable??</u>				
<u>Deterministic??</u>				
<u>Episodic??</u>				
<u>Static??</u>				
<u>Discrete??</u>				
<u>Single-agent??</u>				

The environment type largely determines the agent design

The real world is (of course) partially observable, stochastic, sequential, dynamic, continuous, multi-agent

Environment types

	Crossword	Robo-selector	Poker	Taxi
<u>Observable??</u>	Yes	Partly	Partly	Partly
<u>Deterministic??</u>	Yes	No	No	No
<u>Episodic??</u>	No	Yes	No	No
<u>Static??</u>	Yes	No	Yes	No
<u>Discrete??</u>	Yes	No	Yes	No
<u>Single-agent??</u>	Yes	Yes	No	No

The environment type largely determines the agent design

The real world is (of course) partially observable, stochastic, sequential, dynamic, continuous, multi-agent

Problem types

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Deterministic, fully observable \implies single-state problem

Agent knows exactly which state it will be in; solution is a sequence

Non-observable \implies conformant problem

Agent may have no idea where it is; solution (if any) is a sequence

Nondeterministic and/or partially observable \implies contingency problem

percepts provide new information about current state

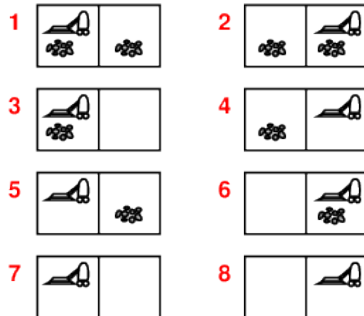
solution is a contingent plan or a policy

often interleave search, execution

Unknown state space \implies exploration problem (“online”)

Example: vacuum world

Single-state, start in #5. Solution??



Example: vacuum world

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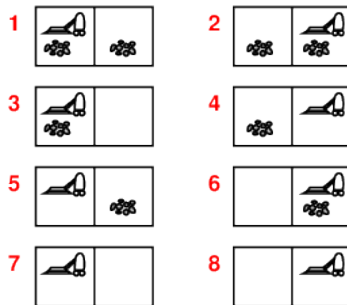
Single-state, start in #5. Solution??

[*Right*, *Suck*]

Conformant

start in {1, 2, 3, 4, 5, 6, 7, 8}

e.g., *Right* goes to {2, 4, 6, 8}. Solution??



Example: vacuum world

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Single-state, start in #5. Solution??
[*Right, Suck*]

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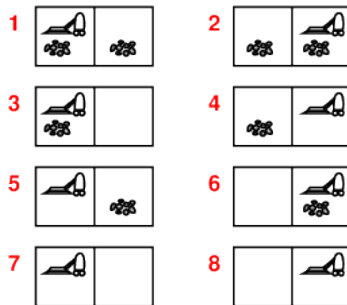
e.g., *Right* goes to {2, 4, 6, 8}. Solution??
[*Right, Suck, Left, Suck*]

Contingency, start in #5

Murphy's Law: *Suck* can dirty a clean carpet

Local sensing: dirt, location only.

Solution??



Example: vacuum world

Single-state, start in #5. Solution??
[*Right, Suck*]

Conformant

start in {1, 2, 3, 4, 5, 6, 7, 8}

e.g., *Right* goes to {2, 4, 6, 8}. Solution??
[*Right, Suck, Left, Suck*]

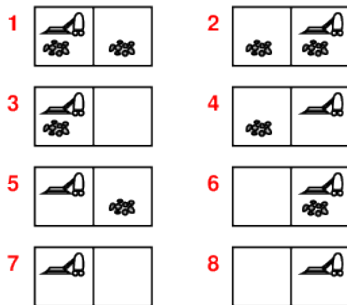
Contingency, start in #5

Murphy's Law: *Suck* can dirty a clean carpet

Local sensing: dirt, location only.

Solution??

[*Right, if dirt then Suck*]



Problem-solving agents

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Restricted form of general agent: **Goal based agents**

- formulate a goal and a problem given the current state
- search for a solution
- execute the solution **ignoring** perceptions

Note: this is **offline** problem solving; solution executed “eyes closed.”

Online problem solving involves acting without complete knowledge.

Problem-solving agents

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```
function Simple-Problem-Solving-Agent(percept) returns an action
  static: seq, an action sequence, initially empty
           state, some description of the current world state
           goal, a goal, initially null
           problem, a problem formulation

  state ← Update-State(state, percept)
  if seq is empty then
    goal ← Formulate-Goal(state)
    problem ← Formulate-Problem(state, goal)
    seq ← Search(problem)
  action ← First(seq)
  seq ← Rest(seq)
  return action
```

An example: Traveling in Romania

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Example (Holidays in Romania)

On holiday in Romania; currently in Arad.

Flight leaves tomorrow from Bucharest

Formulate goal:

- be in Bucharest

Formulate problem:

- states: various cities

- actions: drive between cities

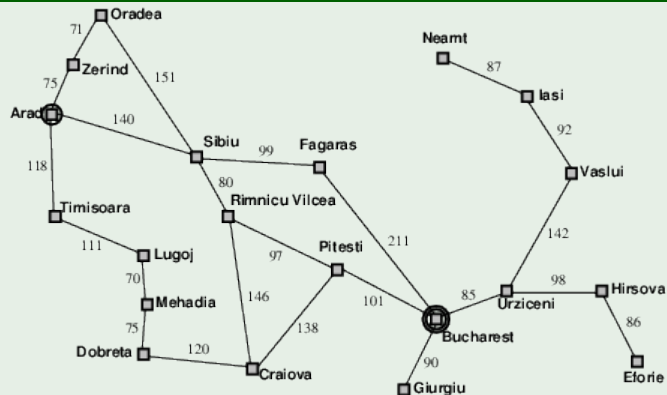
Find solution:

- sequence of cities, e.g., Arad, Sibiu, Fagaras, Bucharest

An example: Traveling in Romania

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Example (Holidays in Romania)



Single-state problem formulation

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A **problem** is defined by four items:

initial state e.g., “at Arad”

successor function $S(x)$ = set of action–state pairs

e.g., $S(A) = \{ \langle \text{Arad} \rightarrow \text{Zerind}, \text{Zerind} \rangle, \dots \}$

goal test, can be

explicit, e.g., $x = \text{“at Bucharest”}$

implicit, e.g., $\text{NoDirt}(x)$

path cost (additive)

e.g., sum of distances, number of actions executed, etc.

$c(x, a, y)$ is the **step cost**, assumed to be ≥ 0

A **solution** is a sequence of actions

leading from the initial state to a goal state

Selecting a state space

Real world is absurdly complex

⇒ state space must be **abstracted** for problem solving

(Abstract) state = set of real states

(Abstract) action = complex combination of real actions

e.g., “Arad → Zerind” represents a complex set
of possible routes, detours, rest stops, etc.

For guaranteed realizability, **any** real state “in Arad”

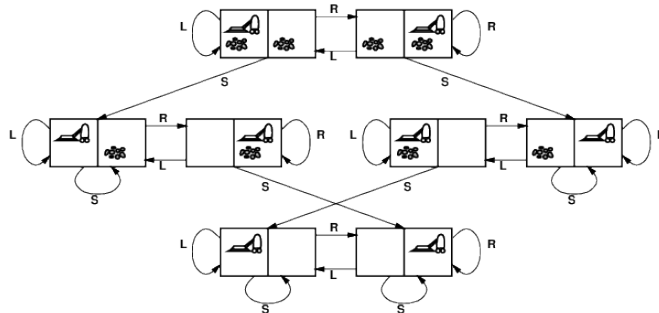
must get to **some** real state “in Zerind”

(Abstract) solution =

set of real paths that are solutions in the real world

Each abstract action should be “easier” than the original problem!

Example: vacuum world state space graph



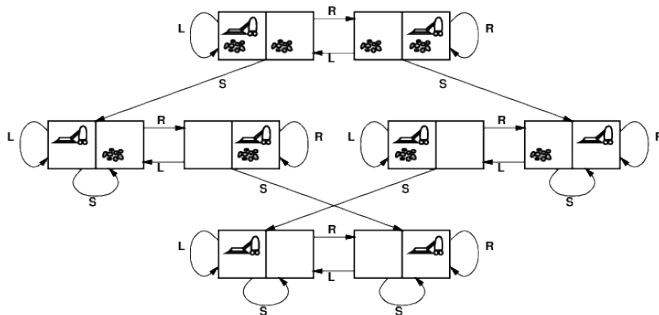
states??:

actions??:

goal test??:

path cost??:

Example: vacuum world state space graph



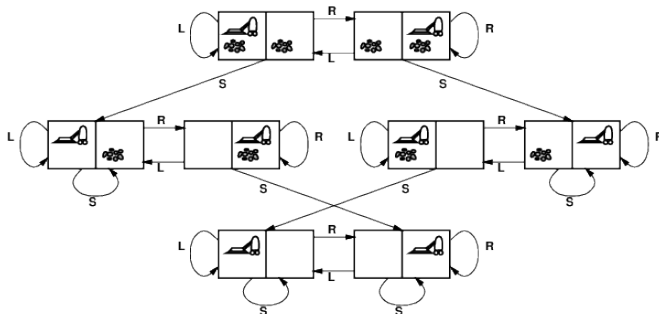
states??: discrete dirt and robot locations (ignore dirt amounts)

actions??:

goal test??:

path cost??:

Example: vacuum world state space graph



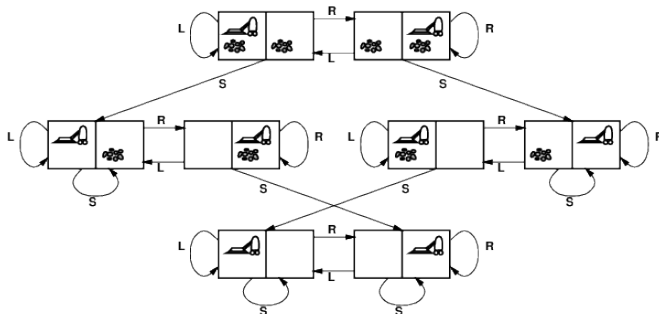
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goal test??:

path cost??:

Example: vacuum world state space graph



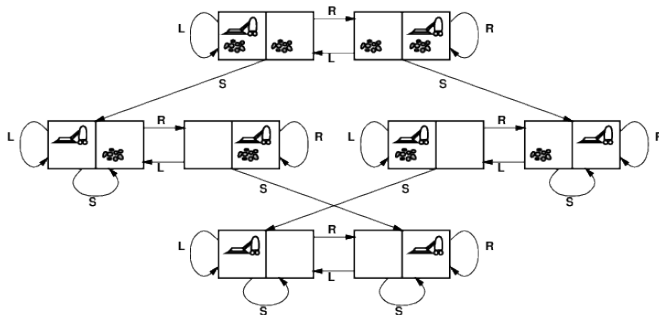
states??: discrete dirt and robot locations (ignore dirt **amounts**)

actions??: *Left, Right, Suck, NoOp*

goal test??: no dirt

path cost??:

Example: vacuum world state space graph



states??: discrete dirt and robot locations (ignore dirt amounts)

actions??: *Left*, *Right*, *Suck*, *NoOp*

goal test??: no dirt

path cost??: 1 per action (0 for *NoOp*)

Example: The 8-puzzle

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7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

states??:

actions??:

goal test??:

path cost??:

Example: The 8-puzzle

7	2	4
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Goal State

states??: integer locations of tiles (ignore intermediate positions)

actions??:

goal test??:

path cost??:

Example: The 8-puzzle

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Start State

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Goal State

states??: integer locations of tiles (ignore intermediate positions)

actions??: move blank left, right, up, down

goal test??:

path cost??:

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Example: The 8-puzzle

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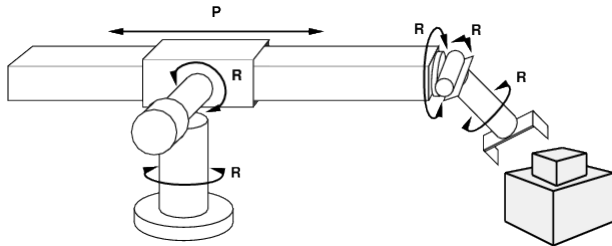
goal test??: given goal state

path cost??: 1 per move

[Note: optimal solution of n -Puzzle family is NP-hard]

Example: robotic assembly

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states??: real-valued coordinates of robot joint angles
parts of the object to be assembled

actions??: continuous motions of robot joints

goal test??: complete assembly **with no robot included!**

path cost??: time to execute

Tree search algorithm

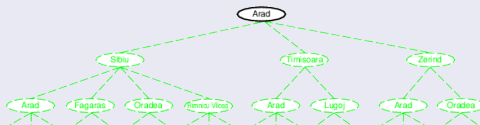
Basic idea:

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 no candidates for expansion then return failure
    choose a leaf node for expansion according to strategy
    if node contains a goal state then return the solution
    else add successor nodes to the search tree (expansion)
  end
```

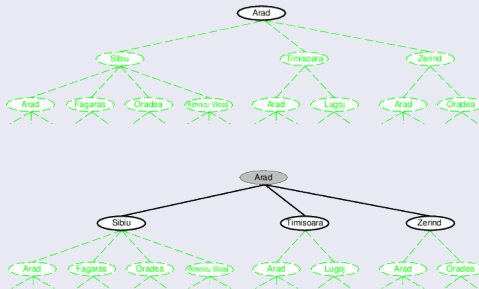
Tree search example

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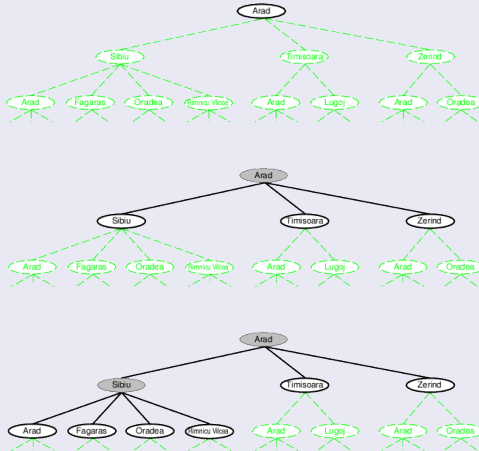
Tree search example

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Tree search example

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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**, **action**, **children**, **depth**, **path cost** (i.e., $g(x)$)

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

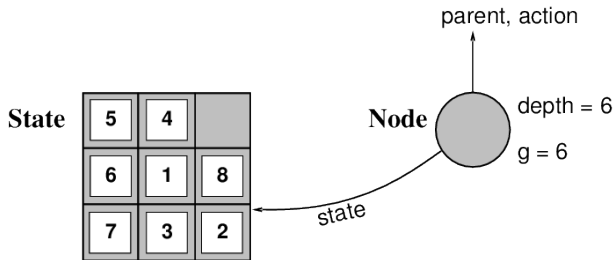
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States do not have parents, actions, children, depth, or path cost!



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, frontier ) returns a solution, or failure
  frontier  $\leftarrow$  Insert( Make-Node( problem.Initial-State ) )
  while not IsEmpty( frontier ) do
    node  $\leftarrow$  Pop( frontier )
    if problem.Goal-Test( node.State ) then return node
    frontier  $\leftarrow$  InsertAll( Expand( node, problem ) )
  end loop
  return failure
```

Implementation: expand nodes

```
function Expand(node, problem) returns a set of nodes
    successors  $\leftarrow$  the empty set
    for each action, result in Successor-Fn(problem, node.State)
    do
        s  $\leftarrow$  a new Node
        s.Parent-Node  $\leftarrow$  node;
        s.Action  $\leftarrow$  action;
        s.State  $\leftarrow$  result
        s.Path-Cost  $\leftarrow$  node.Path-Cost +
            Step-Cost(node.State, action, result)
        s.Depth  $\leftarrow$  node.Depth + 1
        add s to successors
    return successors
```

Search strategies

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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

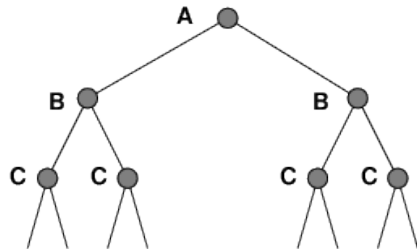
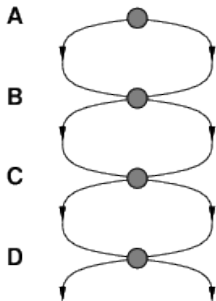
b —maximum branching factor of the search tree

d —depth of the least-cost solution

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

Repeated states

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



Graph search

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```
function Graph-Search(problem, frontier) returns a solution, or failure
  explored  $\leftarrow$  an empty set
  frontier  $\leftarrow$  Insert(Make-Node(problem.Initial-State))
  while not IsEmty(frontier) do
    node  $\leftarrow$  Pop(frontier)
    if problem.Goal-Test(node.State) then return node
    if node.State is not in explored then
      add node.State to explored
      frontier  $\leftarrow$  InsertAll(Expand(node, problem))
    end if
  end loop
  return failure
```