



Artificial and Computational Intelligence

AIMLCZG557

Contributors & Designers of document content: Cluster Course Faculty Team

M2:: Problem Solving Agent using Search

Presented by Faculty Name BITS Email ID

Pilani Campus

Artificial and Computational Intelligence

Disclaimer and Acknowledgement



- Few content for these slides may have been obtained from prescribed books and various other source on the Internet
- I hereby acknowledge all the contributors for their material and inputs and gratefully acknowledge people others who made their course materials freely available online.
- I have provided source information wherever necessary
- This is not a full fledged reading materials. Students are requested to refer to the textbook w.r.t detailed content of the presentation deck that is expected to be shared over e-learning portal - taxilla.
- I have added and modified the content to suit the requirements of the class dynamics & live session's lecture delivery flow for presentation
- Slide Source / Preparation / Review:
- From BITS Pilani WILP: Prof.Raja vadhana, Prof. Indumathi, Prof.Sangeetha
- From BITS Oncampus & External: Mr.Santosh GSK

Course Plan

M1	Introduction to Al
M2	Problem Solving Agent using Search
М3	Game Playing
M4	Knowledge Representation using Logics
M5	Probabilistic Representation and Reasoning
M6	Reasoning over time
M7	Ethics in Al

Learning Objective

At the end of this class, students Should be able to:

- 1. Design problem solving agents
- 2. Create search tree for given problem
- 3. Apply uninformed search algorithms to the given problem
- 4. Compare performance of given algorithms in terms of completeness, optimality, time and space complexity
- 5. Differentiate for which scenario appropriate uninformed search technique is suitable and justify



Simple Reflex Agent

function SIMPLE-REFLEX-AGENT(percept) returns an action

persistent: rules, a set of condition-action rules

state←INTERPRET-INPUT(percept)

rule←RULE-MATCH(state, rules)

action ←rule.ACTION

return action

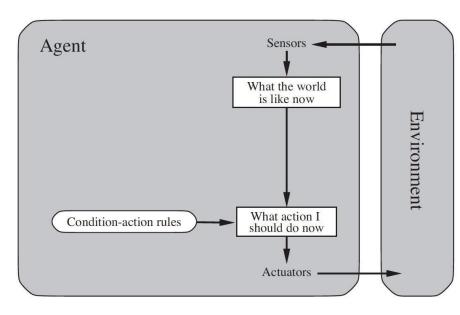
function REFLEX-VACUUM-AGENT([location, status]) returns an action

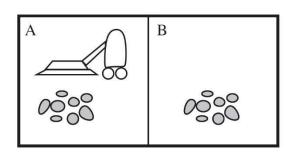
if status = Dirty then return Suck

else if location = A then return Right

else if location = B then return Left

Simple Reflex Agents





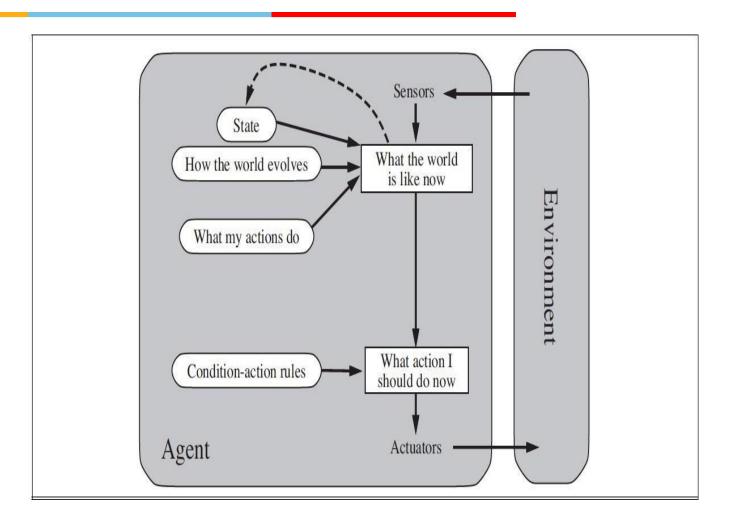


Model based Agent

Simple Reflex Agents



Model Based Agents





Model based Agent

function MODEL-BASED-REFLEX-AGENT(percept) returns an action

persistent: state, the agent's current conception of the world state

transition model, a description of how the next state depends on the current state and action sensor model, a description of how the current world state is reflected in the agent's percepts rules, a set of condition-action rules action, the most recent action, initially none

state←UPDATE-STATE(state, action, percept, transition model, sensor model)

rule←RULE-MATCH(state, rules)

action ←rule.ACTION

return action



Goal based Agent

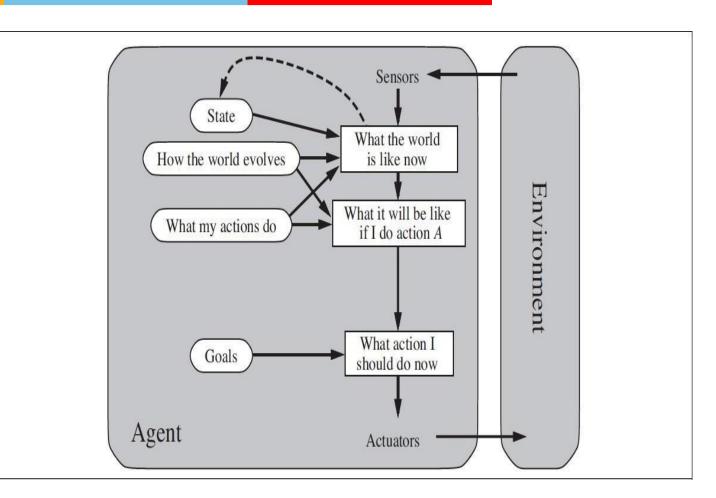
Simple Reflex Agents



Model Based Agents

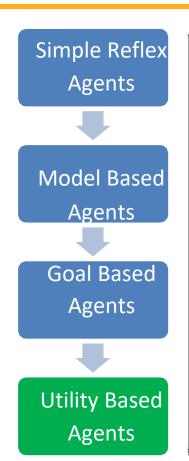


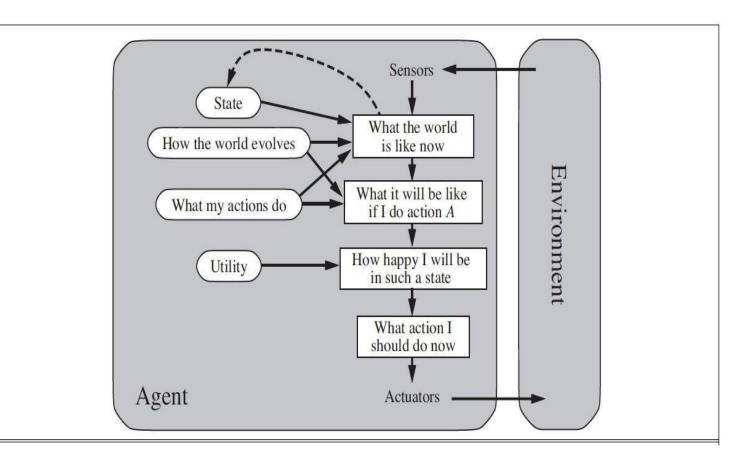
Goal Based Agents





Utility based Agent







Learning Agent

Simple Reflex Agents



Model Based Agents



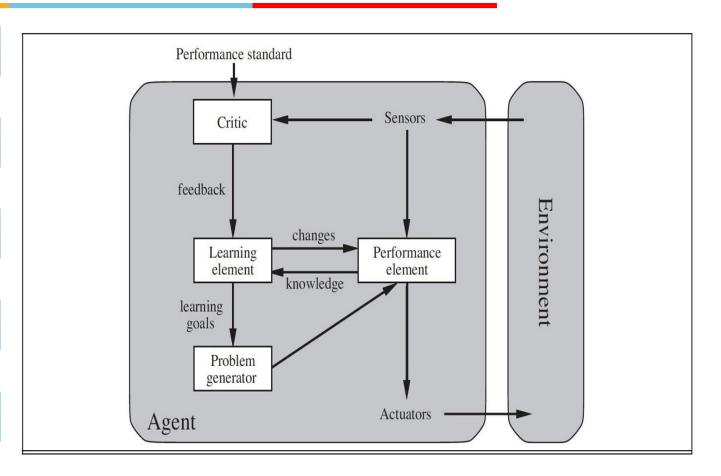
Goal Based Agents



Utility Based Agents



Learning Agents





Performance Element – taking a decision of action based on percepts

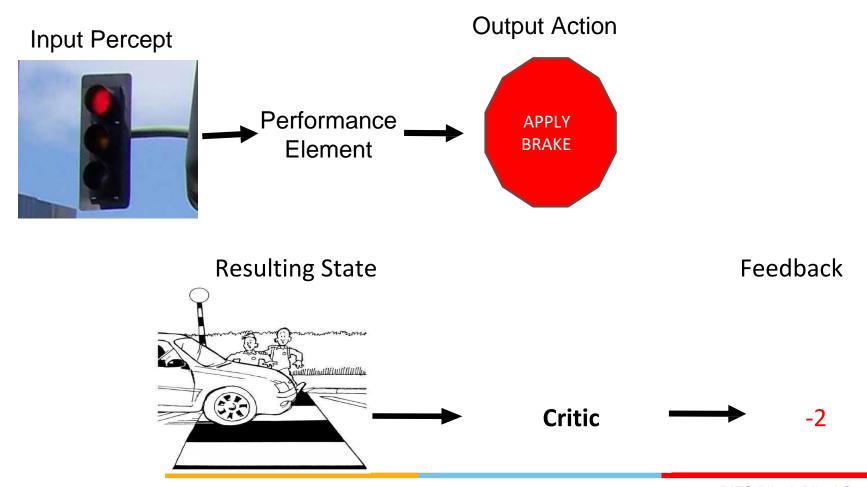
Learning Element – Make the performance element select better actions such that the utility function is optimized

Critic – Provides feedback on the actions taken

Problem Generator – Make the Performance Element select sub-optimal actions such that you would learn from unseen actions



Agents that improve their performance by learning from their own experiences





Input Percept



Possible Actions

Brake
Change Gear to Lower
Change Gear to Higher
Accelerate
Steer left
Steer right

Selected Action

Random

Change Gear to Lower



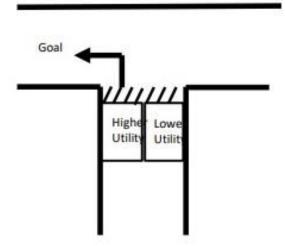
Performance Element – Takes decision on action based on percept

```
f(red \ signal, \ distance) = 15k \ N \ brake

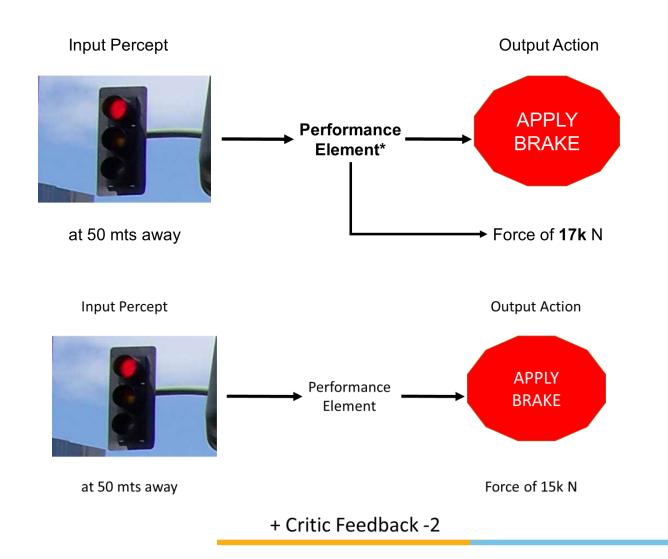
distance = f'(percept \ sequence)

f(percepts, distance, raining)
```

- $f(state_0, actionA) = 0.83,$
- $f(state_0, actionB) = 0.45$



Learning: Supervised Vs Unsupervised Vs Reinforcement



Highe

Utility

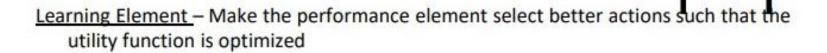
Utilit

Role of Learning

Performance Element - Takes decision on action based on percept

 $f(red\ signal,\ distance) = 15k\ N\ brake$ $distance = f'(percept\ sequence)$ f(percepts, distance, raining)

- $f(state_0, action A) = 0.83,$
- $f(state_0, actionB) = 0.45$



Critic - Provides feedback on the actions taken

<u>Problem Generator</u> – Make the Performance Element select sub-optimal actions such that you would learn from unseen actions

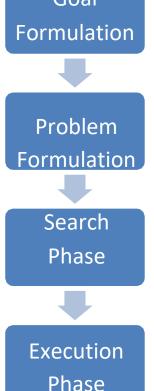
Problem Formulation

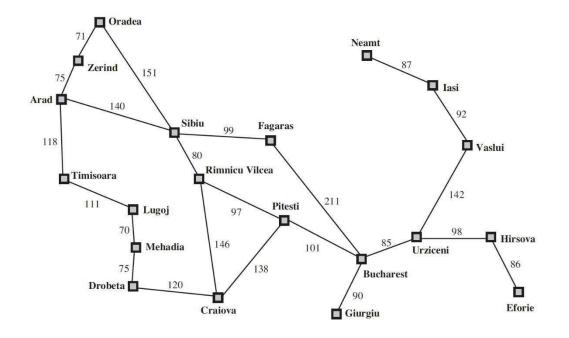


Goal based decision making agents finds sequence of actions that leads to the desirable state.

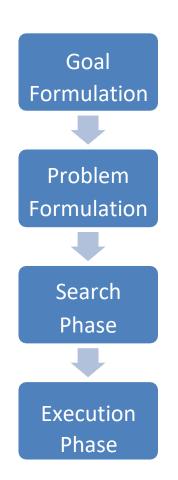
Phases of Solution Search by PSA

Goal Formulation Optimizes the Objective (Local | Global) Limits the Actions



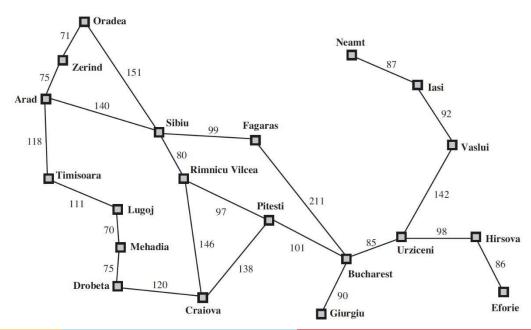




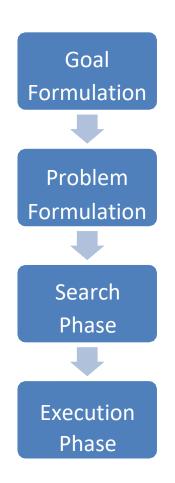


Phases of Solution Search by PSA

State Space Creations [in the path of Goal] Lists the Actions







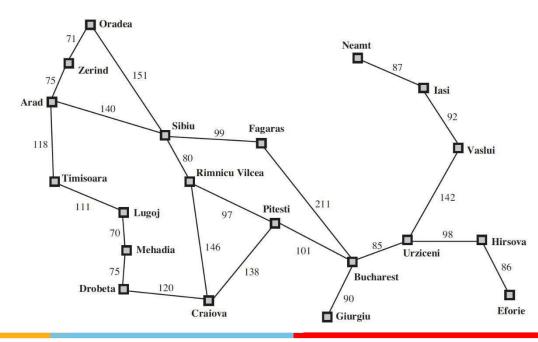
Phases of Solution Search by PSA

Assumptions – Environment :

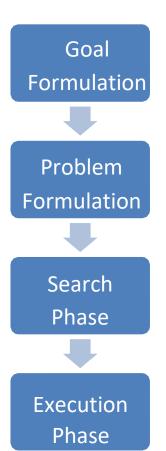
Static

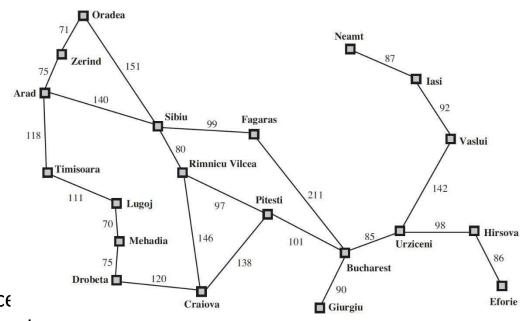
Observable Discrete

Deterministic



Phases of Solution Search







Problem Solving Agents – Problem Formulation

Abstraction Representation
Decide what actions under states to take to achieve a goal

5 Components

Initial State
Possible
Action
| Operators
| Model

Path Cost
| Path Cost

A function that assigns a numeric cost to each path. A path is a series of actions. Each action is given a cost depending on the problem.

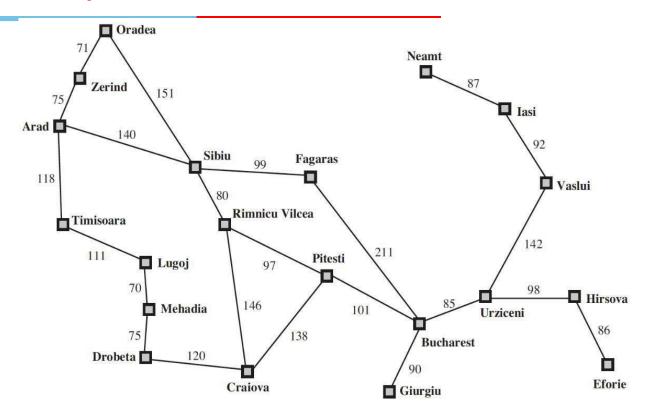
Solution = Path Cost Function + Optimal Solution

Problem Solving Agents - Problem

innovate achieve

lead

Formulation: Book Example



Initial State –E.g., *In(Arad)*

Possible Actions – $ACTIONS(s) \rightarrow \{Go(Sibiu), Go(Timisoara), Go(Zerind)\}$

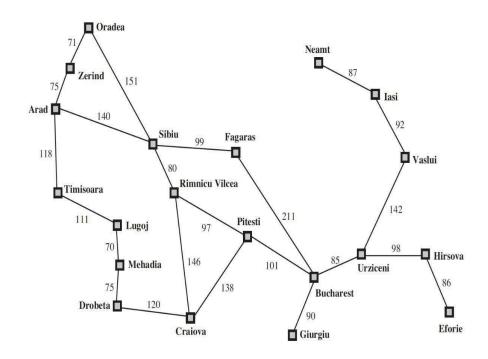
Transition Model – RESULT(In(Arad), Go(Sibiu)) = In(Sibiu)

Goal Test – *IsGoal(In(Bucharest)) = Yes*

Path Cost – cost(In(Arad), go(Sibiu)) = 140 kms

Example Problem Formulation

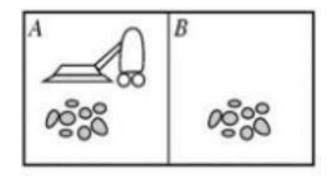
	Travelling Problem
Initial State	Based on the problem
Possible Actions	Take a flight Train Shop
Transition Model/ Successor Function	[A, Go(A->S)] = [S]
Goal Test	Is current = B (destination)
Path Cost	Cost + Time + Quality





Example Problem Formulation

	Vacuum World
Initial State	Any
Possible Actions	[Move Left, Move Right, Suck, NoOps]
Transition Model/ Successor Function	[A, ML] = [B, Dirty] [A, ML] = [B, Clean]
Goal Test	Is all room clean? [A, Clean] [B, Clean]
Path Cost	No of steps in path





Example Problem Formulation

	N-Queen
Initial State	Empty Partial Full
Possible Actions	
Transition Model/	
Successor Function	
Goal Test	
Path Cost	

	0	1	2	3
0			₩	
1	₩			
2				깥
3		₩		

board[r][c]

Path finding Robot

Successor Function Design

1	2	3	4	5	6	0
	8		10	11	12	1
13	14		16	17	18	2
19	20		22	23	24	3
25	26	27			30	4
	32	33		35	36	5
37	38	39	40	41	42	6
0	1	2	3	4	5	•

N-W-E-S

Graph Searching

Graph as state space (node = state, edge = action)

➤ For example, game trees, mazes, ...

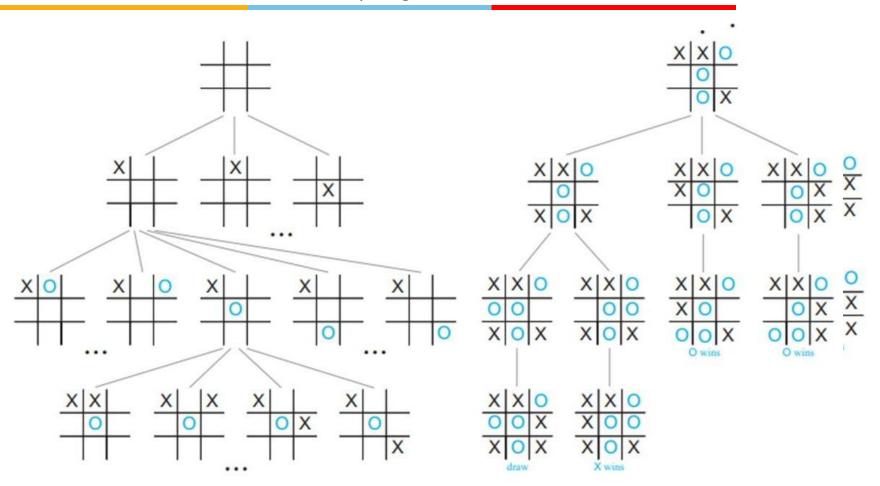


FIGURE 8 Some of the Game Tree for Tic-Tac-Toe.

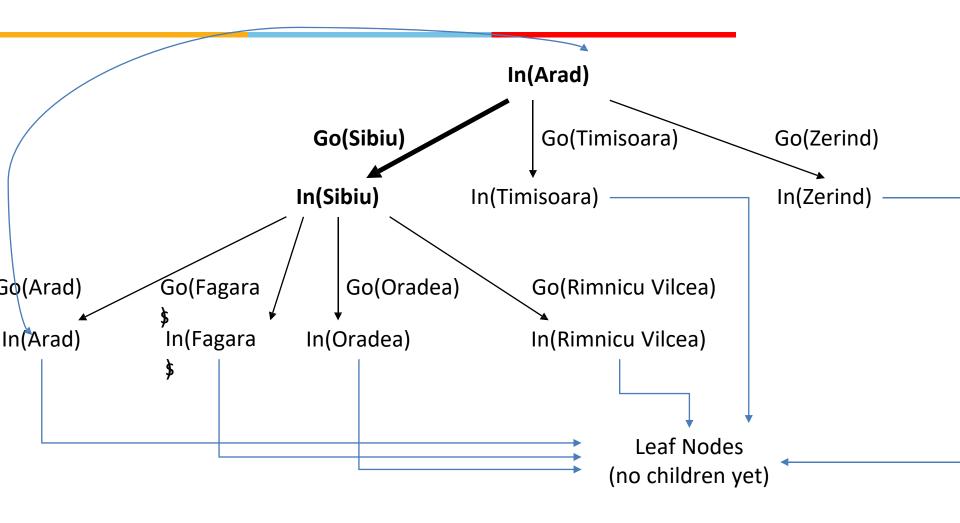
lead



Searching for Solutions

Choosing the current state, testing possible successor function, expanding current state to generate new state is called Traversal. Choice of which state to expand – Search Strategy





Next Class Plan

- Uninformed Search Algorithms
 - ➤ BFS vs DFS An overview
 - > Uniform Cost Search
 - ➤ Iterative Depth First Search
- > Informed Search Algorithms
 - Greedy Best First search
 - ➤ A* Search (Start)

Required Reading: AIMA - Chapter #1, 2, 3.1, 3.2, 3.3

Thank You for all your Attention

Note: Some of the slides are adopted from AIMA TB materials