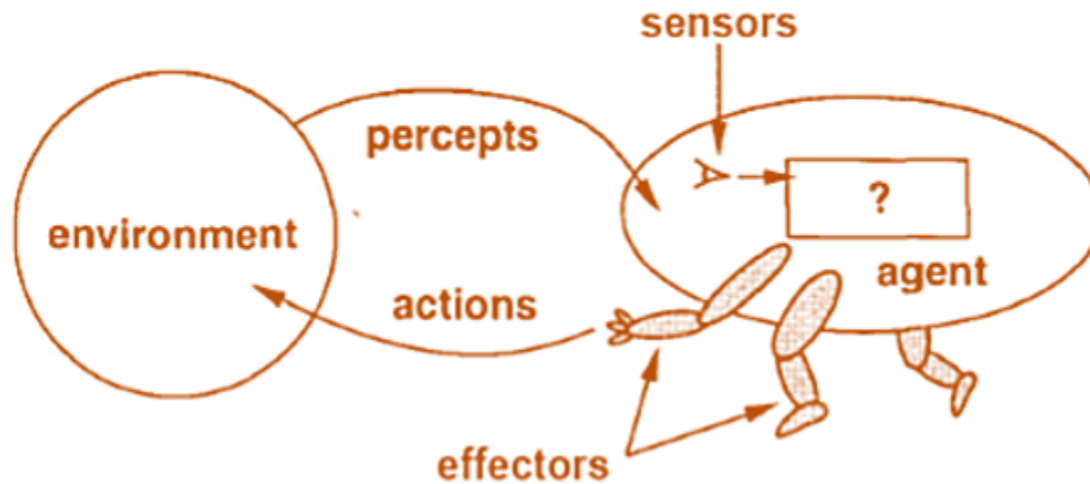


BCSE306

Artificial Intelligence

Module 1: Introduction

Sub topic: **Structure of Agent & its type**



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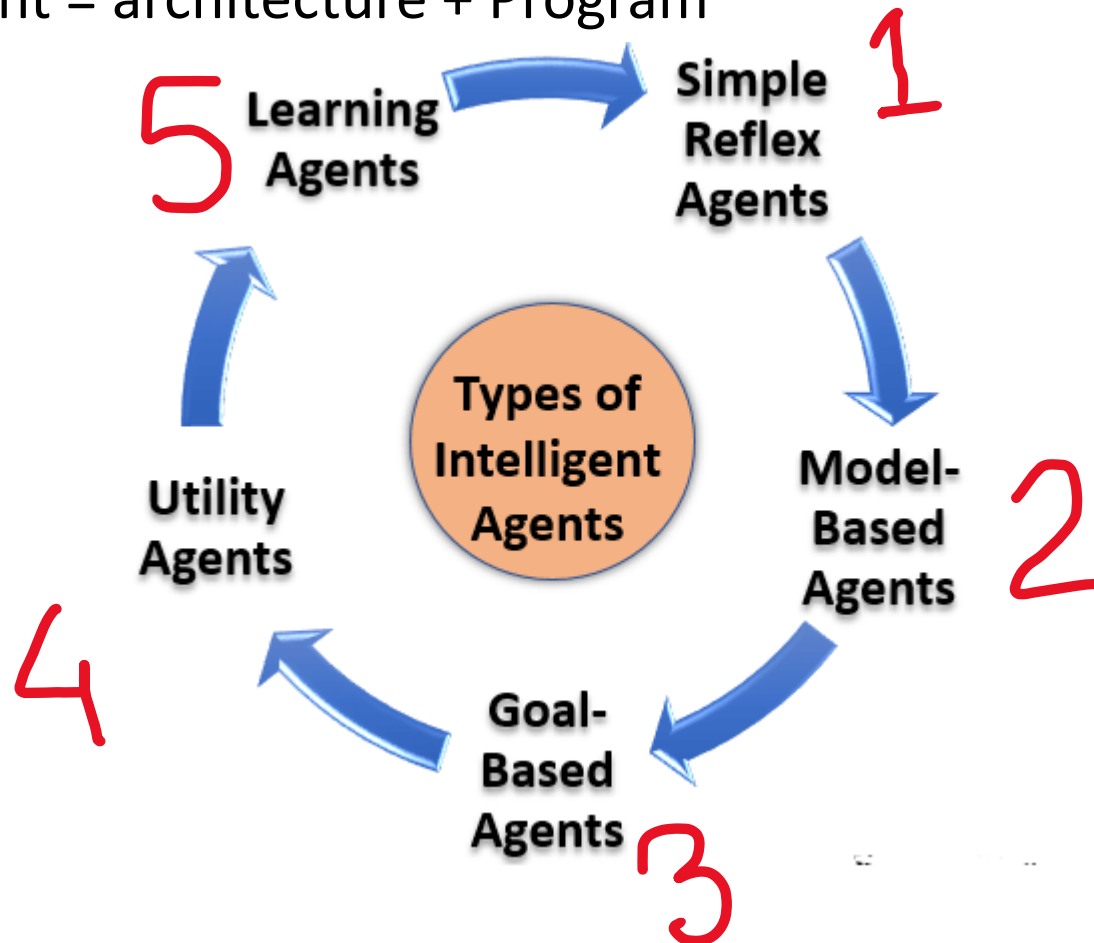
VIT Vellore,

Lecture Outline

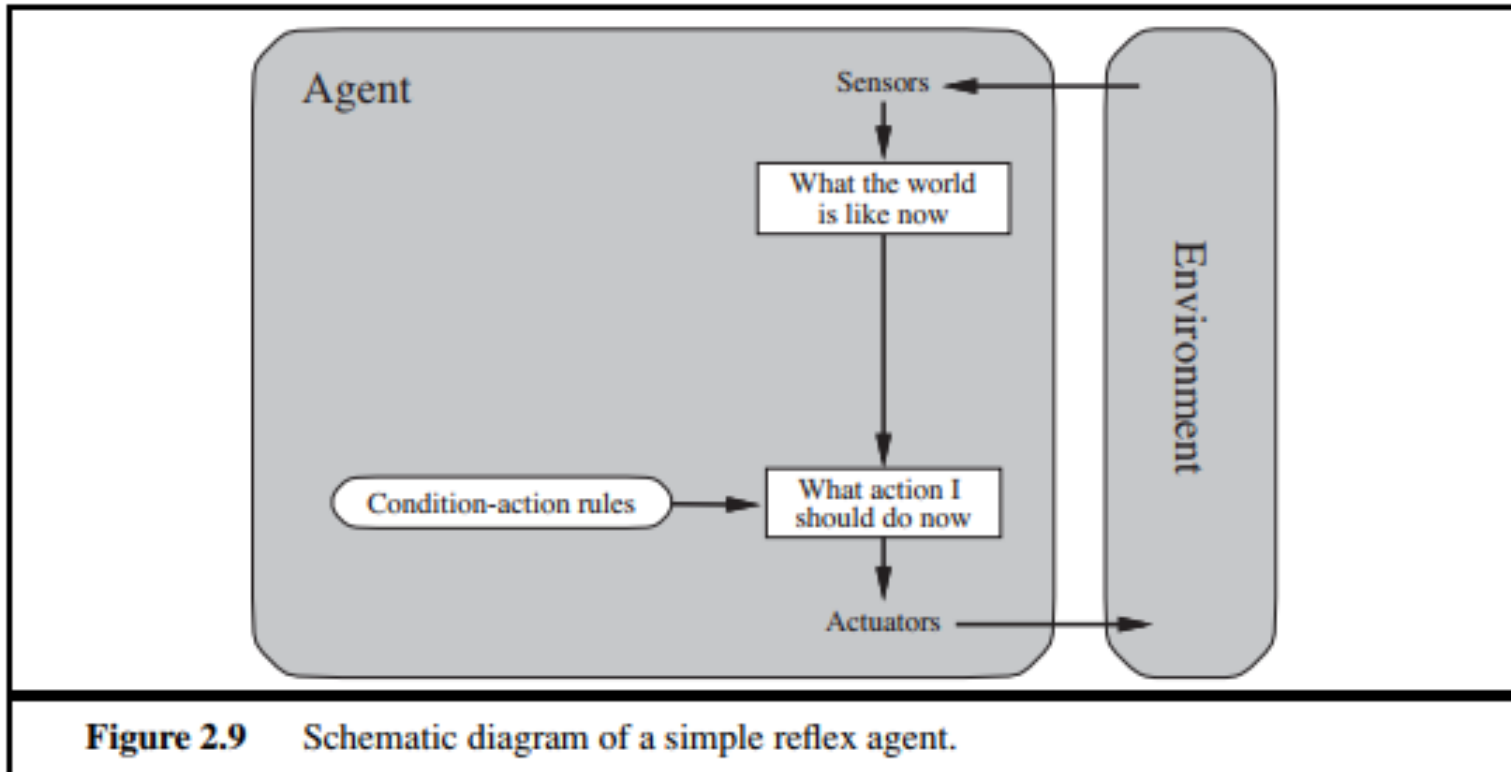
- The structure of Agents
 - Agent = architecture + Program
 - Agent Programs
 - Simple Reflex Agents
 - Model-based reflex agents
 - Goal-based agents
 - Utility based agent
 - Learning based agents

Lecture Outline

- The structure of Agents
 - Agent = architecture + Program



Agent Architecture



Agent Program

```
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
```

Figure 2.10 A simple reflex agent. It acts according to a rule whose condition matches the current state, as defined by the percept.

Structure of Agent

- Till now, we have talked about **agents** by describing **behavior**—the action that is performed after any given sequence of percepts.
- The job of AI is to design an **agent program** that implement the agent function
 - Mapping from percepts to actions
- **Assumption:** this program will run on some sort of computing device with physical sensors and actuators (called as architecture).

agent = architecture + program

Agent Program

agent program is represented using pseudo code. E.g **Table driven Agent**.

- Trivial agent program that keeps track of the **percept sequence** and then uses it to **index into a table of action**.
- The table represents explicitly the agent function that the agent program embodies.
- In order to build a **rational agent**, the designers must **construct a table** that contains the **appropriate action** for every **possible percept sequence**

Table Driven Agent: Agent Program

```
function TABLE-DRIVEN-AGENT(percept) returns an action
  persistent: percepts, a sequence, initially empty
               table, a table of actions, indexed by percept sequences, initially fully specified

  append percept to the end of percepts
  action  $\leftarrow$  LOOKUP(percepts, table)
  return action
```

Figure 2.7 The TABLE-DRIVEN-AGENT program is invoked for each new percept and returns an action each time. It retains the complete percept sequence in memory.

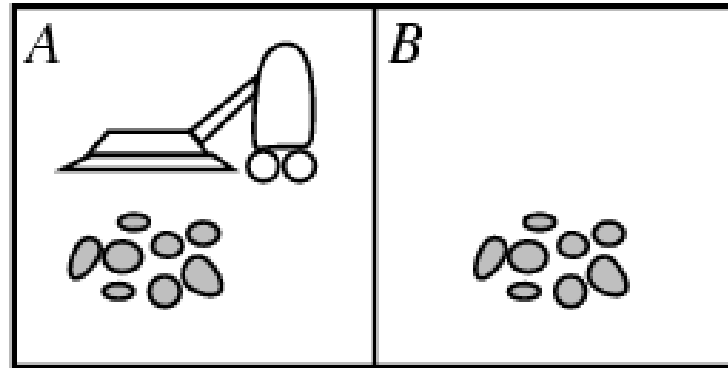
Table Driven Agent: Example

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*



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

Actions: *Left, Right, Suck, NoOp*

Table Driven Agent: Drawbacks

- ▶ Huge table (**Example:** chess requires 10^{150} entries in the lookup table)
- ▶ Take a long time to build the table
- ▶ No autonomy
- ▶ Even with learning, need a long time to learn the table entries

Agent Types

- ▶ Five basic types in order of increasing generality
- 1. Simple reflex agents
- 2. **Model-based reflex agents**
- 3. Goal-based agents
- 4. **Utility-based agents**
- 5. **Learning agent**

Simple Reflex Agents

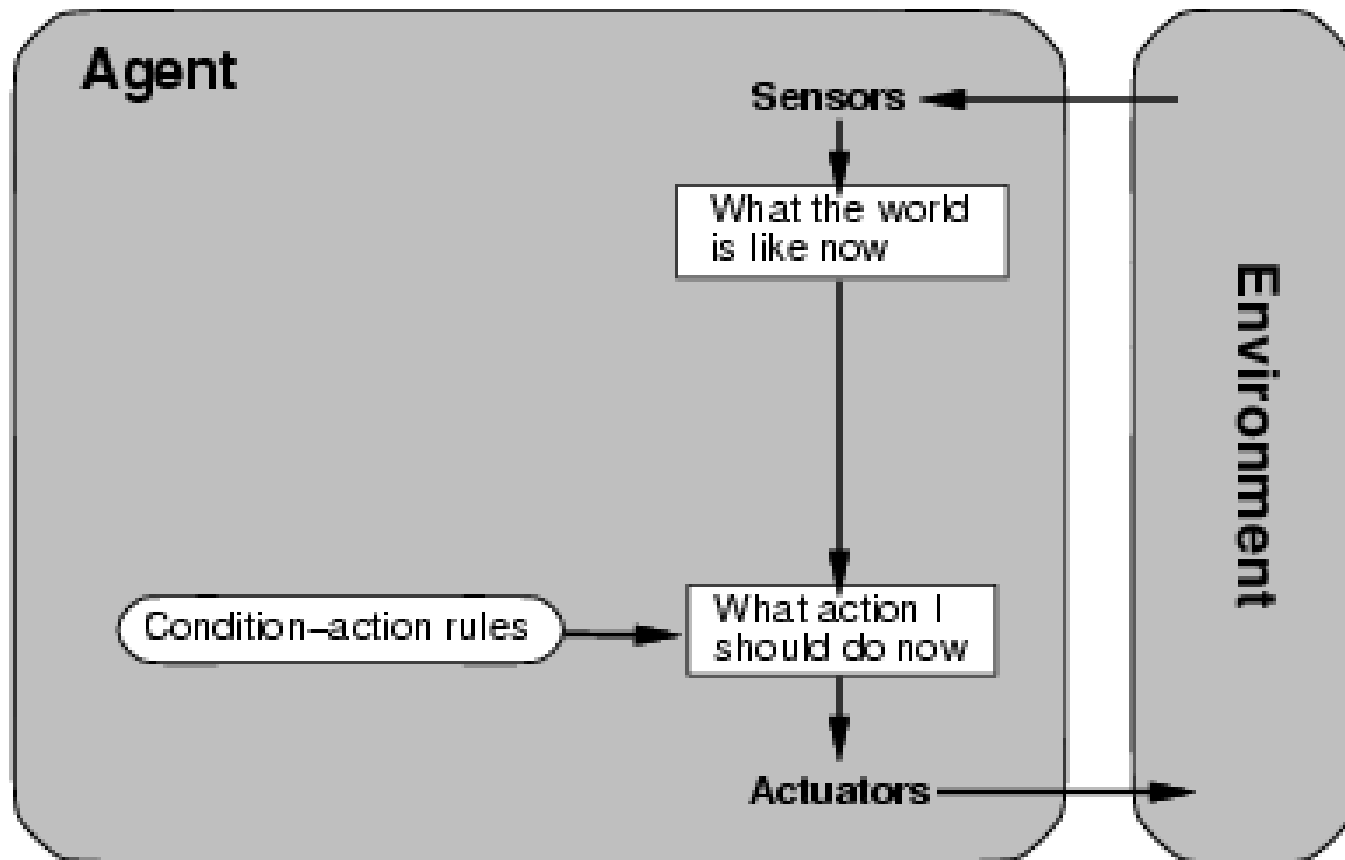
- ▶ It select actions on the basis of the current percept, ignoring the rest of the percept history

VACUUM-AGENT

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>[l3, Dirty]</i>	<i>Suck</i>
<i>[A, Clean], [A, Clean]</i>	<i>Right</i>
<i>[A, Clean], [A, Dirty]</i>	<i>Suck</i>
•	•
<i>[A, Clean], [A, Clean], [A, Clean]</i>	<i>Right</i>
<i>[A, Clean], [A, Clean], [A, Dirty]</i>	<i>Suck</i>
⋮	•

Works on condition–action rule !

Simple Reflex Agents



Simple Reflex Agents

```
function SIMPLE-REFLEX-AGENT(percept) returns an action  
persistent: rules, a set of condition–action rules  
  
  state  $\leftarrow$  INTERPRET-INPUT(percept)  
  rule  $\leftarrow$  RULE-MATCH(state, rules)  
  action  $\leftarrow$  rule.ACTION  
  return action
```

- The INTERPRET-INPUT function generates an abstracted description of the current state from the percept
- RULE-MATCH function returns the first rule in the set of rules that matches the given state description

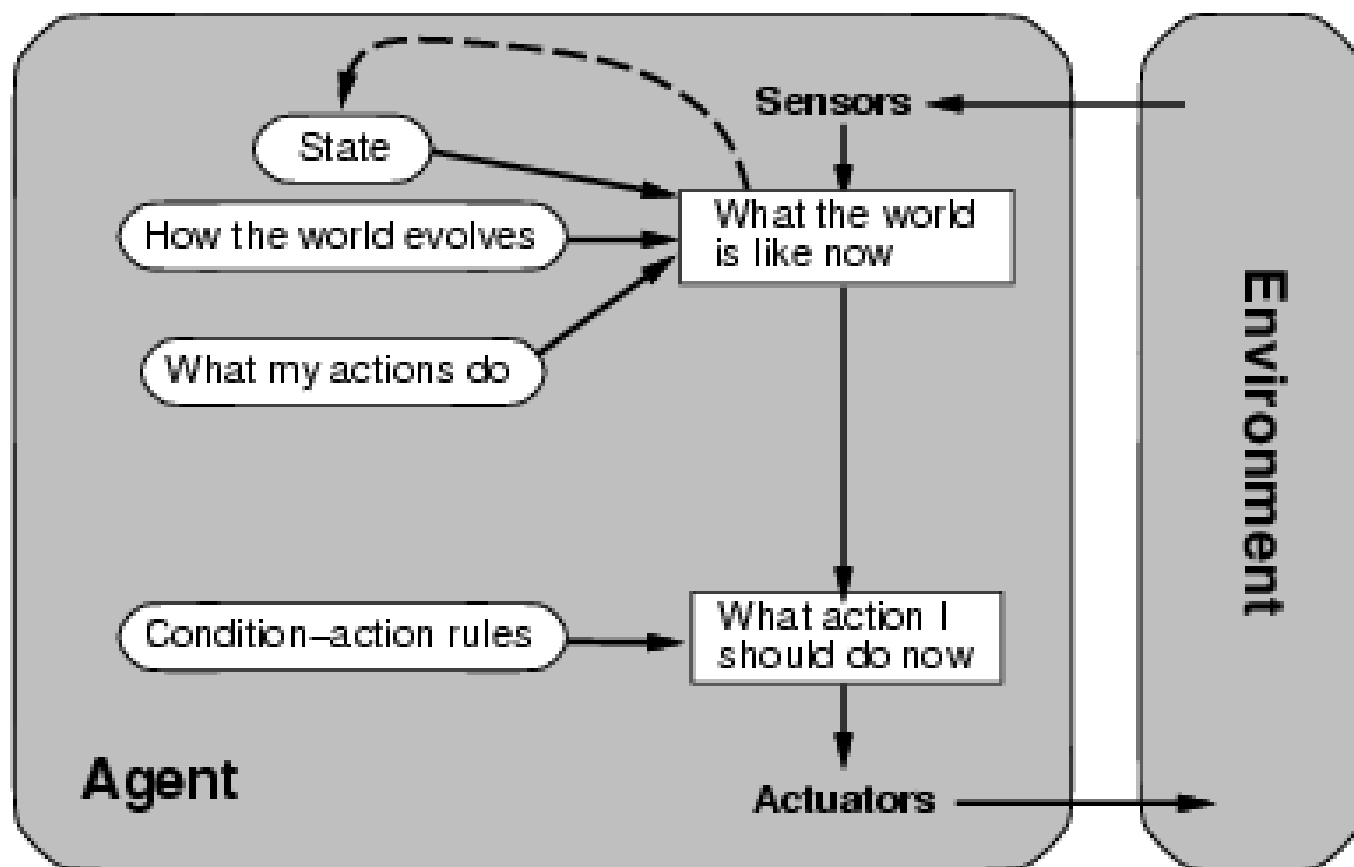
Simple Reflex Agents

- Advantages:
 - Easy to implement
 - Uses much less memory than the table-driven agent
- Disadvantages:
 - Will only work correctly if the environment is fully observable
 - Infinite loops

Model-based Reflex Agents

- ▶ The agent should maintain some sort of *internal state that depends on the percept history* and thereby *reflects* at least some of the unobserved aspect of the current state
- ▶ Updating this internal state information as requires two kinds of knowledge to be encoded in the agent program
 - ▶ Information about how the world evolves independently of the agent
 - ▶ Information about how the agent's own actions affect the world

Model-based Reflex Agents



Model-based Reflex Agents

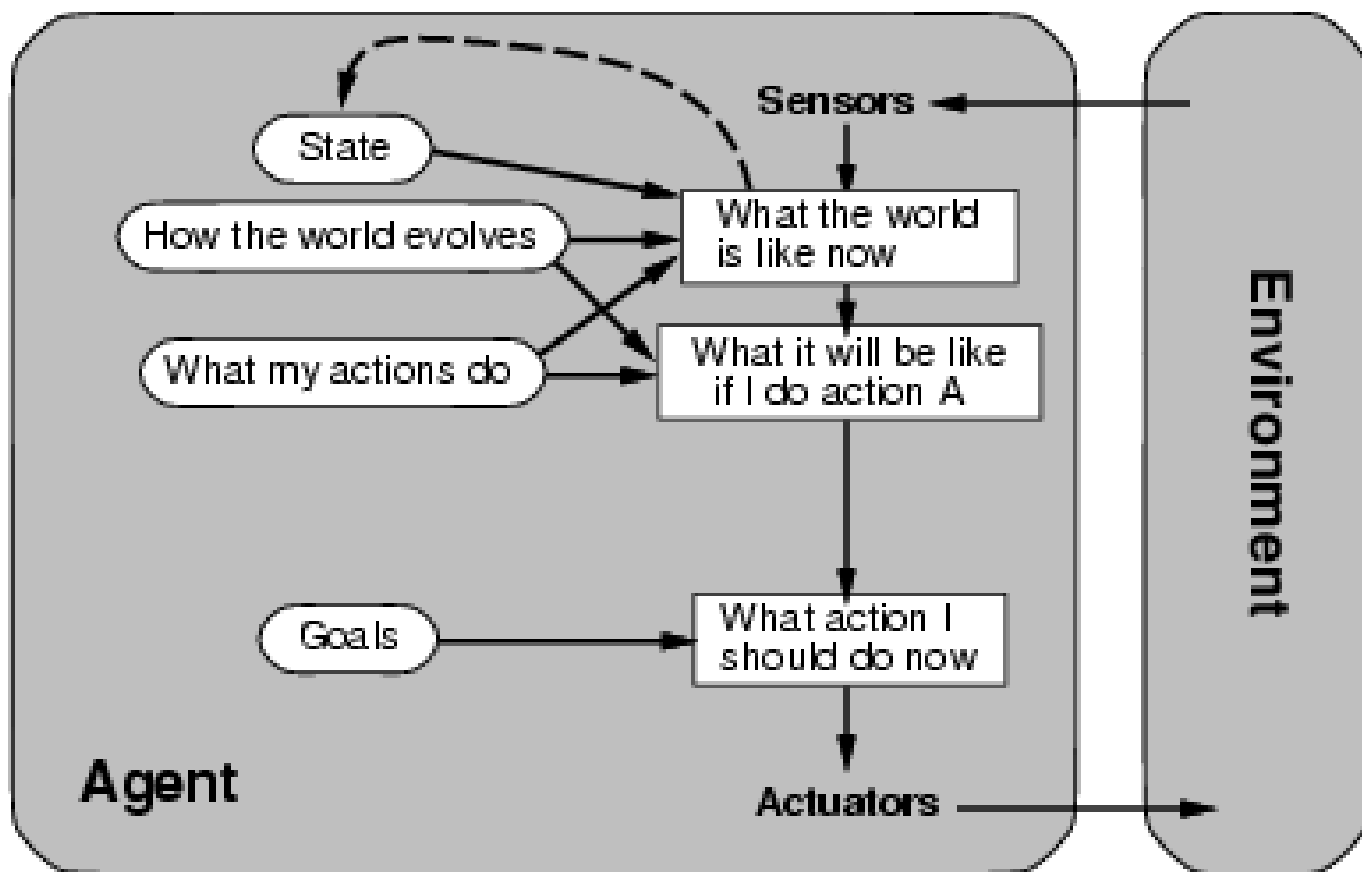
```
function REFLEX-AGENT-WITH-STATE(percept) returns an action  
  static: state, a description of the current world state  
          rules, a set of condition-action rules  
          action, the most recent action, initially none  
  
  state ← UPDATE-STATE(state, action, percept)  
  rule ← RULE-MATCH(state, rules)  
  action ← RULE-ACTION[rule]  
  return action
```

- ▶ The above said state information is implemented in simple Boolean circuits or in complete scientific theories called **model** of the world

Goal-based Agents

- ▶ Goal information guides agent's actions (looks to the future)
- ▶ Sometimes achieving goal is **simple** eg. from a single action
- ▶ Other times, goal requires **reasoning** about long sequences of actions
- ▶ **Flexible**: simply reprogram the agent by changing goals

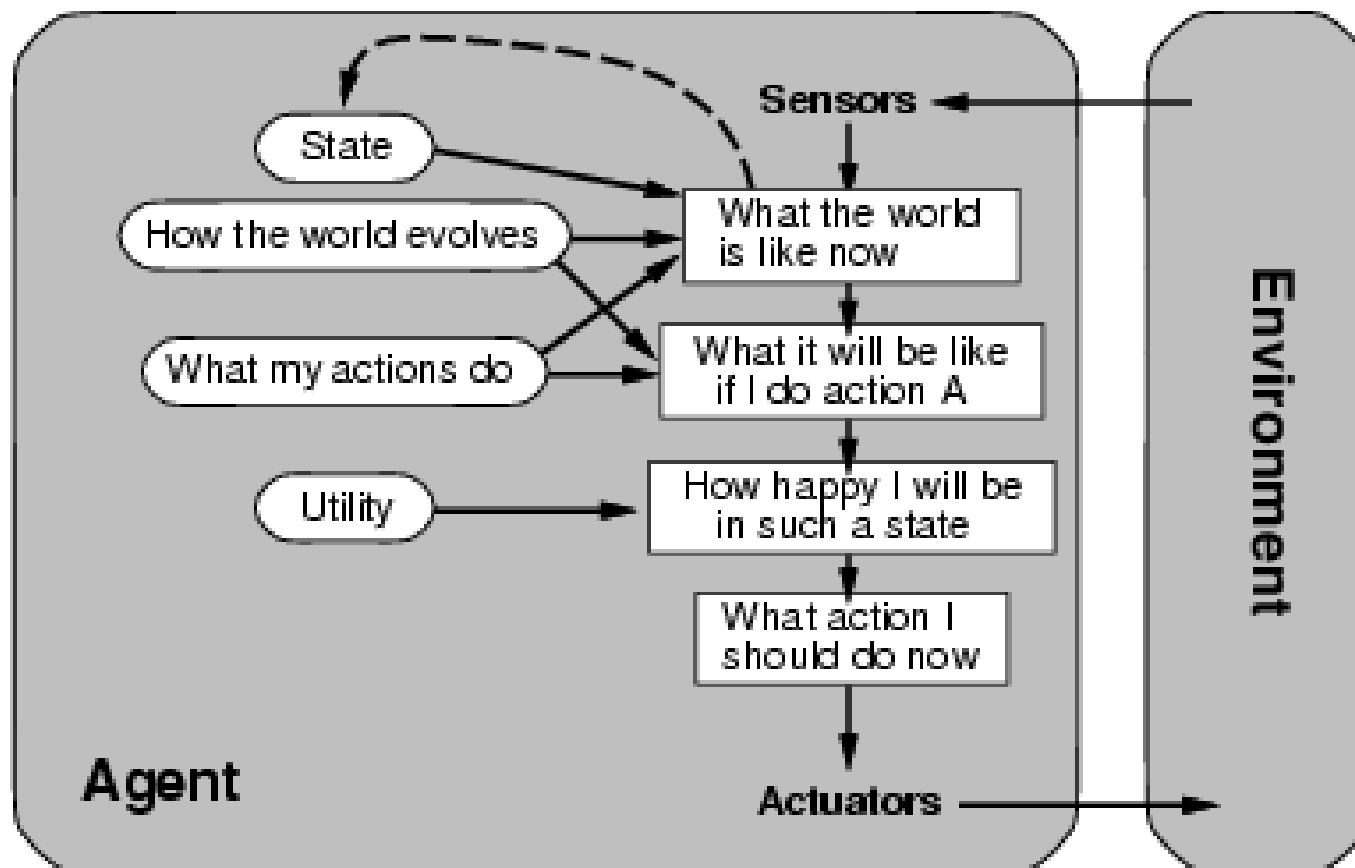
Goal-based Agents



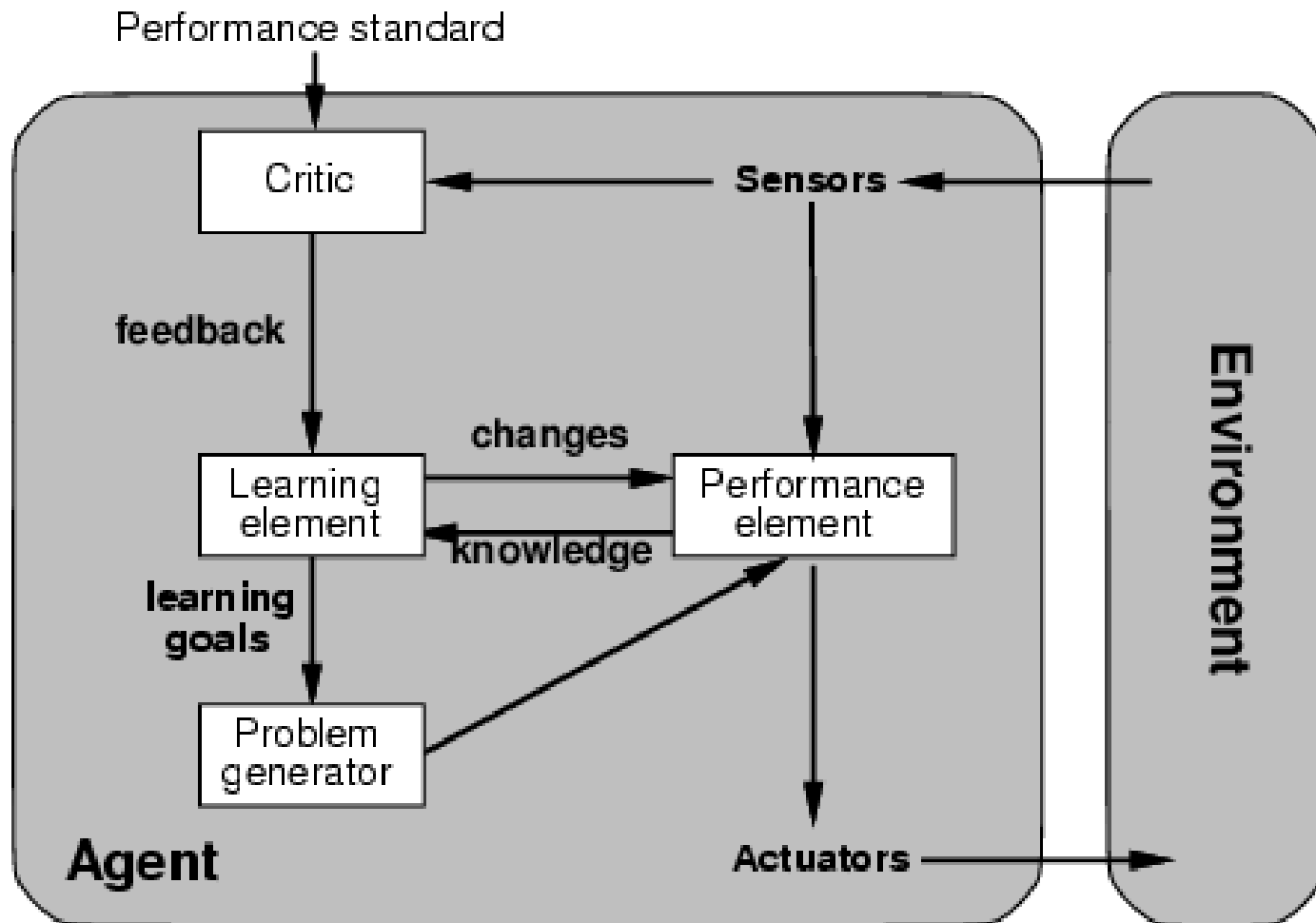
Utility-based Agents

- ▶ What if there are many paths to the goal?
- ▶ Utility measures which states are preferable to other state
- ▶ Maps state to real number (utility)

Utility-based Agents



Learning Agents



Learning Agents

- ▶ **Learning element** is responsible for making improvements
- ▶ **Performance element** is responsible for selecting external actions.
- ▶ **Critic** provides feedback on how the agent is doing and determines how the performance element should be modified to do better in the future.
- ▶ **Problem generator** is responsible for suggesting actions that will lead to new and informative experiences.



AI based Systems

Expert Systems –
**Apply expert
knowledge to
difficult,
real world problems.**

Knowledge Based Systems –
Make domain knowledge explicit

AI Programs-
**Exhibit intelligent behavior by skillful
application of heuristics**



Project Work

► Sample projects / ideas :

1. <http://web.stanford.edu/class/cs221/2017/project-list.html>
2. <https://www.cse.iitb.ac.in/~ananddhoot/projects.html>
3. [http://www.cs.cornell.edu/courses/cs478/2001sp/mllinks/interesting ai demos and project.htm](http://www.cs.cornell.edu/courses/cs478/2001sp/mllinks/interesting_ai_demos_and_project.htm)
4. <http://web.cs.iastate.edu/~cs572/projects.html>
5. <https://web.stanford.edu/class/cs221/2018/project-list.html>
6. <http://cs229.stanford.edu/projects2011.html>
7. web.stanford.edu/class/cs224w/projects.html

► Note:

- **Plagiarism is not permitted**, kindly don't copy someone's project and pass it off as yours. If found you will get **0 marks** for the whole project.

Research Work/Survey

▶ Areas:

- ▶ Geo Sensing/ Geo-forecasting/ Geo-Science
- ▶ Image Reorganization
- ▶ NLP
- ▶ Vehicular Networks
- ▶ Intelligent Transportation Systems
- ▶ Autonomous Vehicles
- ▶ Data Science (Text Analysis, Sentiment Analysis Clustering)
- ▶ Bio Technology / Medical Bio Engineering/ Bio-Informatics
- ▶ Robotics
- ▶ Gaming
- ▶ Pervasive Computing
- ▶ ...