

CHAPTER : TWO

Intelligent Agent

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- Types of Agent
- Rational Agents
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- Types of Environments

Intelligent Agent

- I want to build a robot that will
 - Clean my house
 - Information filtering agents
 - Cook when I don't want to
 - Wash my clothes
 - Cut my hair
 - Fix my car (or take it to be fixed)
 - Take a note when I am in a meeting
 - Handle my emails

i.e. do the things that I don't feel like doing...

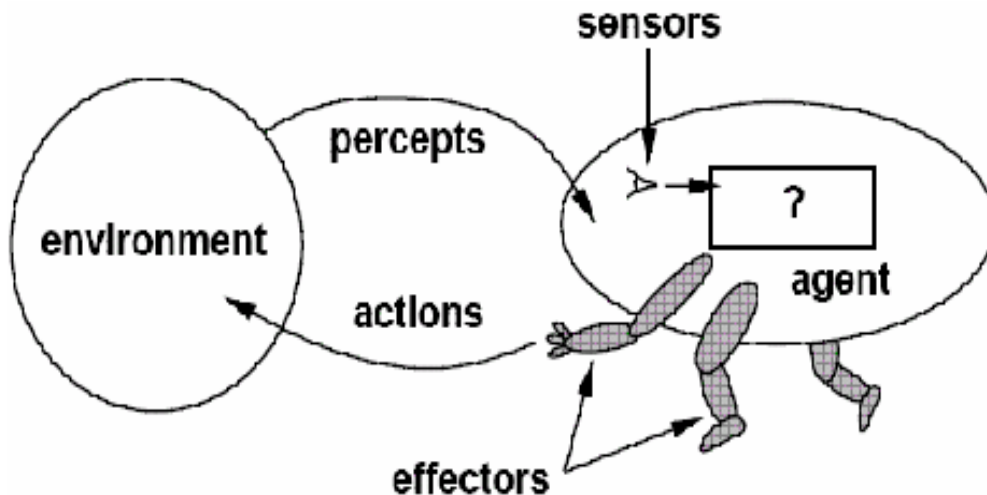
- AI is the science of building machines (agents) that act rationally with respect to a goal.

Types of Intelligent Agents

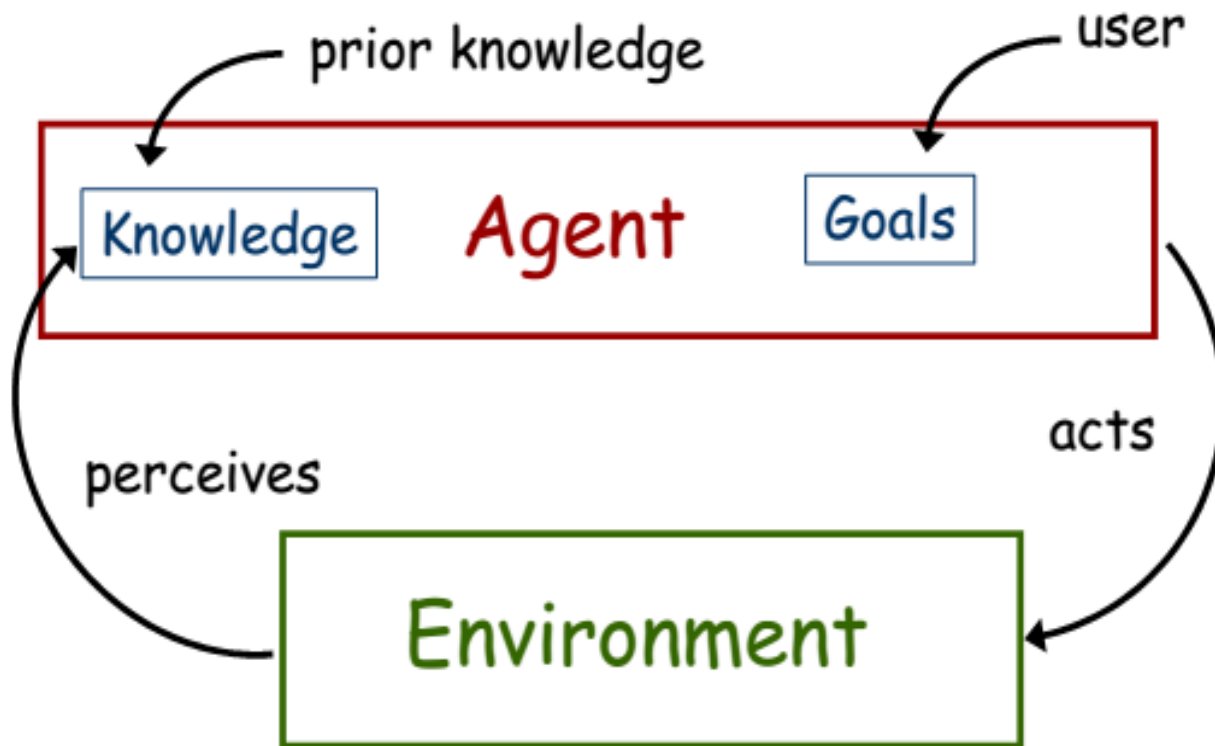
- Software agents:
 - Also called a softbot (software robot)
 - It is an agent that interacts with a software environment by issuing commands and interpreting the environments feedback.
 - E.g. mail handling agent, information filtering agent
- Physical agents
 - are robots that operates in the physical world and can perceive and manipulate objects in that world

Agent

- **Agent** is something that perceives/observes its environment through **SENSORS** and acts upon that environment through **EFFECTORS**.
- The agent is assumed to exist in an environment in which it perceives and acts
- An agent is rational/sensible since it does the right thing to achieve the specified goal.



- **Agent = architecture + program**



- Require more flexible interaction with the environment, the ability to modify one's goals, knowledge that be applied flexibly to different situations

Agent

| | Human beings | Agents |
|--|--------------------|---|
| Sensors | Eyes, Ears, Nose | Cameras, Scanners, Mic, infrared range finders |
| Effectors <i>Handwritten: 8x [20+5</i> | Hands, Legs, Mouth | Various Motors (artificial hand, artificial leg), Speakers, Radio |

Examples of agents in different types of applications

| Agent type | Percepts | Actions | Goals | Environment |
|---------------------------------|-------------------------------------|---|-----------------------------------|--------------------------------|
| Medical diagnosis system | Symptoms, patient's answers | Questions, tests, treatments | Healthy patients, minimize costs | Patient, hospital |
| Interactive English tutor | Typed words, questions, suggestions | Write exercises, suggestions, corrections | Maximize student's score on exams | Set of students, materials |
| Part-picking robot | Pixels of varying intensity | Pick up parts and sort into bins | Place parts in correct bins | Conveyor belts with parts |
| Satellite image analysis system | Pixels intensity, color | Print a categorization of scene | Correct categorization | Images from orbiting satellite |
| Refinery controller | Temperature, pressure readings | Open, close valves; adjust temperature | Maximize purity, yield, safety | Refinery |

Rationality/Reasonableness vs. Omniscience/Awareness/knowledge

- Rational agent acts so as to achieve one's goals, given one's beliefs (one that does the right thing).
 - What does right thing mean? one that will cause the agent to be most successful and is expected to maximize goal achievement, given the available information
- An Omniscient/all-knowing agent knows the actual outcome of its actions, and can act accordingly, but in reality omniscience is impossible.
- Rational agents take action with expected success, where as omniscient agent take action with 100% sure of its success
- Is human beings Omniscient or Rational agent?

Example

- You are walking along the road to **shewaber**; You see an old friend across the street. There is no traffic.
- So, being rational, you start to cross the street.
- Mean while a big banner falls off from above and before you finish crossing the road, you are flattened.

Were you irrational to cross the street?

- This points out that rationality is concerned with expected success, given what has been perceived.
 - Crossing the street was rational, because most of the time, the crossing would be successful, and there was no way you could have foreseen the falling banner.
 - The EXAMPLE shows that **we can not blame an agent for failing to take into account something it could not perceive.** Or for failing to take an action that it is incapable of taking.

Rational agent

- In summary what is rational at any given point depends on four things.
 - **Perception/sensitivity**: Everything that the agent has perceived so far concerning the **current scenario** in the environment
 - **Knowledge**: What an agent **already knows** about the environment
 - **Action**: The actions that the agent can perform back to the environment
 - **Performance measure**: The performance measure that defines **degrees of success** of the agent.
- Generally, rationality refers to “**doing the right thing**”.
 - ✓ So that one **key property** of Intelligent Agents is **being a rational agent**, because rational agents means doing the right thing.
- Therefore in **designing an intelligent agent**, one has to remember **PEAS** (**P**erformance, **E**nvironment, **A**ctuators, **S**ensors) framework.

Action

- **Example 1: PEAS description for Automated Taxi Driver.**
 - a) Performance Measure
 - ✓ How much it is safe, fast, legal, comfortable trip, maximize profits, etc.
 - b) Environment
 - ✓ The roads, pedestrians, customers, other traffic.
 - c) Actuators
 - ✓ The steering wheel, accelerator, brake, signal, horn, etc.
 - d) Sensors
 - ✓ The cameras, sonar, speedometer, GPS, odometer, engine sensors, keyboard, etc.

- **Example 2:** PEAS description of **Part-Sorting Robot**
 - a) Performance Measure
 - ✓ Percentage of parts in correct bins
 - b) Environment
 - ✓ Conveyor belt with parts, bins
 - c) Actuators
 - ✓ Robotic arm
 - d) Sensors
 - ✓ Camera, joint angle sensors

- **Example 3: PEAS description of Spam Filter**
 - a) Performance Measure
 - ✓ Minimizing false positives, false negatives
 - b) Environment
 - ✓ A user's email account
 - c) Actuators
 - ✓ Mark as spam, delete, etc.
 - d) Sensors
 - ✓ Incoming messages, other information about user's account

Performance measure

- How do we decide whether an agent is successful or not?
 - Establish a standard of what it means to be successful in an environment and use it to measure the performance.
 - A rational agent should do whatever action is expected to maximize its performance measure, on the basis of the evidence provided by the percept sequence and whatever built-in knowledge the agent has.
- What is the performance measure for “**crossing the road**”?
- What about “**Chess Playing**”?

Assignment

- Consider the need to design a “taxi driver agent” that serves in Debre markos city;
 - Identify what to perceive, actions to take, the environment it interacts with?
 - Identify **sensors, effectors, goals, environment** and **performance measure** that should be integrated for the agent to be successful in its operation?

Designing an agent/Agent Function

- An **agent function** maps the **percept histories** to **actions**.
- An **agent program** is the one that runs on the **physical architecture** to produce the agent function.
- **Agent = architecture + program**
- Architecture
 - Runs the programs
 - Makes the percept from the sensors **available** to the programs
 - **Feeds** the program's **action choices** to the effectors
- Programs
 - **Accepts percept from an environment and generates actions**
 - Before designing an agent program, we need to know the **possible percept and actions**
 - By enabling a learning mechanism, the agent could have a degree of autonomy/independence, such that it can reason and take decision

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Types of Environments

1. Fully Observable (vs. Partially Observable)

- The agent's sensors give it access to the complete state of the environment at each point in time.

2. Deterministic (vs. Stochastic)

- The next state of the environment is completely determined by the current state and the agent's action.

3. Episodic (vs. Sequential)

- The agent's experience is divided into atomic “episodes,” and the choice of action in each episode depends only on the episode itself.

4. Static (vs. Dynamic)

- The environment is unchanged while an agent is deliberating.
- **Semi dynamic**:- The environment does not change with the passage of time, but the agent's performance score does.

5. Discrete (vs. Continuous)

- The environment provides a fixed number of distinct percepts, actions, and environment states.
- Time can also evolve in a discrete or continuous fashion.

6. **Single Agent (vs. Multi-agent)**

- An agent operating **by itself** in an environment

7. **Known (vs. Unknown)**

- The **agent knows the rules of the environment**

Case Examples for Task Environment Types

- Example 1: Task environment types of three agents: *Chess*, *Taxi with Clock* and *Taxi without Clock*

| | | | |
|-------------------------|-----------|-----------|----|
| Fully Observable | Yes | Yes | No |
| Deterministic | Strategic | Strategic | No |
| Episodic | No | No | No |
| Static | Semi | Yes | No |
| Discrete | Yes | Yes | No |
| Single Agent | No | No | No |

- **Example 2:** Task Environments types of four agents:
Chess, *Poker*, *Image Analysis* and *Butler Robot*

| Task Env. | Observable | Agents | Episodic | Static | Discrete |
|----------------|------------|--------|------------|---------|------------|
| Chess | Fully | Multi | Sequential | static | discrete |
| Poker | Partially | Multi | Sequential | static | discrete |
| Image analysis | Fully | Single | Episodic | static | continuous |
| Butler Robot | Partially | Single | Sequential | dynamic | continuous |

Environment Types

Below are lists of properties of a number of familiar environments

| Problems | Observable | Deterministic | Episodic | Static | Discrete |
|----------------------------|------------|---------------|----------|--------|----------|
| Crossword Puzzle | Yes | Yes | No | Yes | Yes |
| Part-picking/cutting robot | No | No | Yes | No | No |
| Web shopping program | No | No | No | No | Yes |
| Tutor | No | No | No | Yes | Yes |
| Medical Diagnosis | No | No | No | No | No |
| Taxi driving | No | No | No | No | No |

•Hardest case: a environment that is inaccessible, sequential, non-deterministic, dynamic, continuous.

Agents acting in an environment



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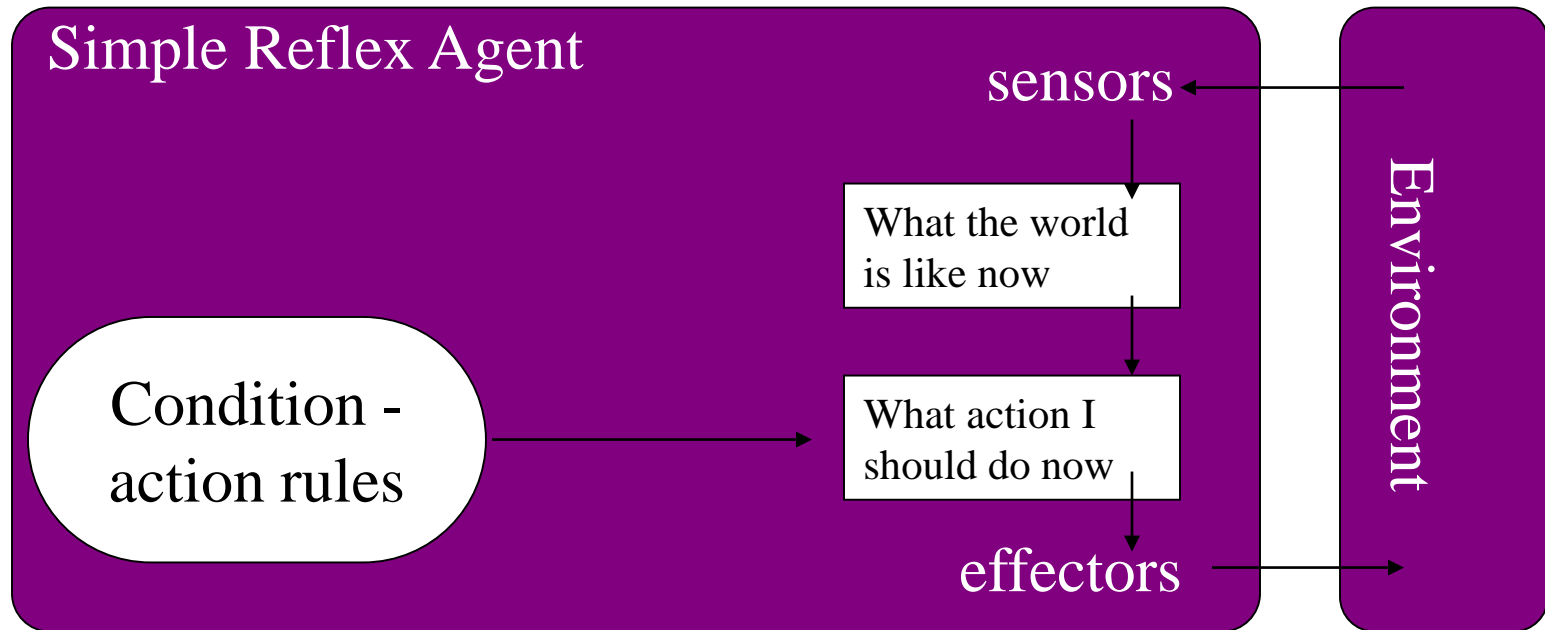
Types **Hierarchy of Agent Types**

- Let's have a closer look on the following five agent categories:-
 - I. Simple Reflex Agents**
 - II. Model-Based Reflex Agents**
 - III. Goal-Based Agents**
 - IV. Utility-Based Agents**
 - V. Learning Agents**

I. Simple Reflex Agents

- Select action on the basis of current percept, ignoring all past percepts
 - It uses just *condition-action rules*
 - ✓ The rules are like the form “if ... then ...”
 - ✓ Efficient but have narrow range of applicability
 - ✓ Because knowledge sometimes cannot be stated explicitly
 - ✓ Work only
 - If the environment is fully observable
- If car-in-front-is breaking then initiate-braking.
 - Blinking when something approaches the eye.

Figure: Structure of a simple reflex agent



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

static: *rules*, a set of condition-action rules

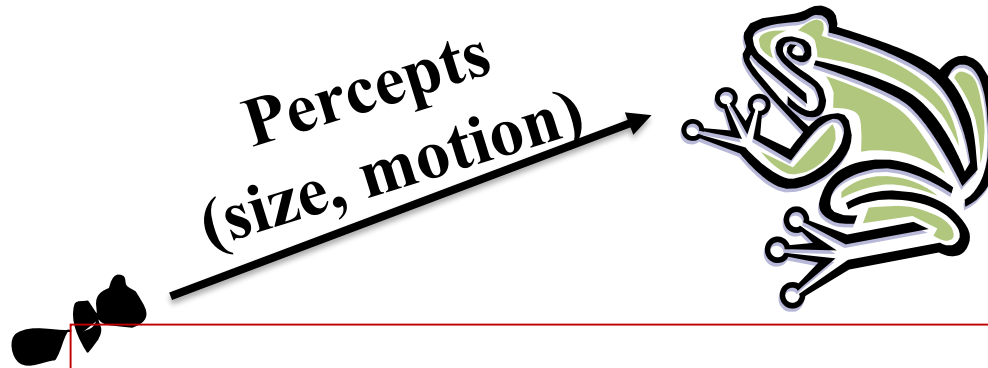
state \leftarrow INTERPRET-INPUT (*percept*)

rule \leftarrow RULE-MATCH (*state*, *rules*)

action \leftarrow RULE-ACTION [*rule*]

return *action*

A Simple Reflex Agent in Nature



RULES

- (1) If small moving object, then activate SNAP
 - (2) If large moving object,
then activate AVOID and inhibit SNAP
- ELSE (not moving) then NOOP

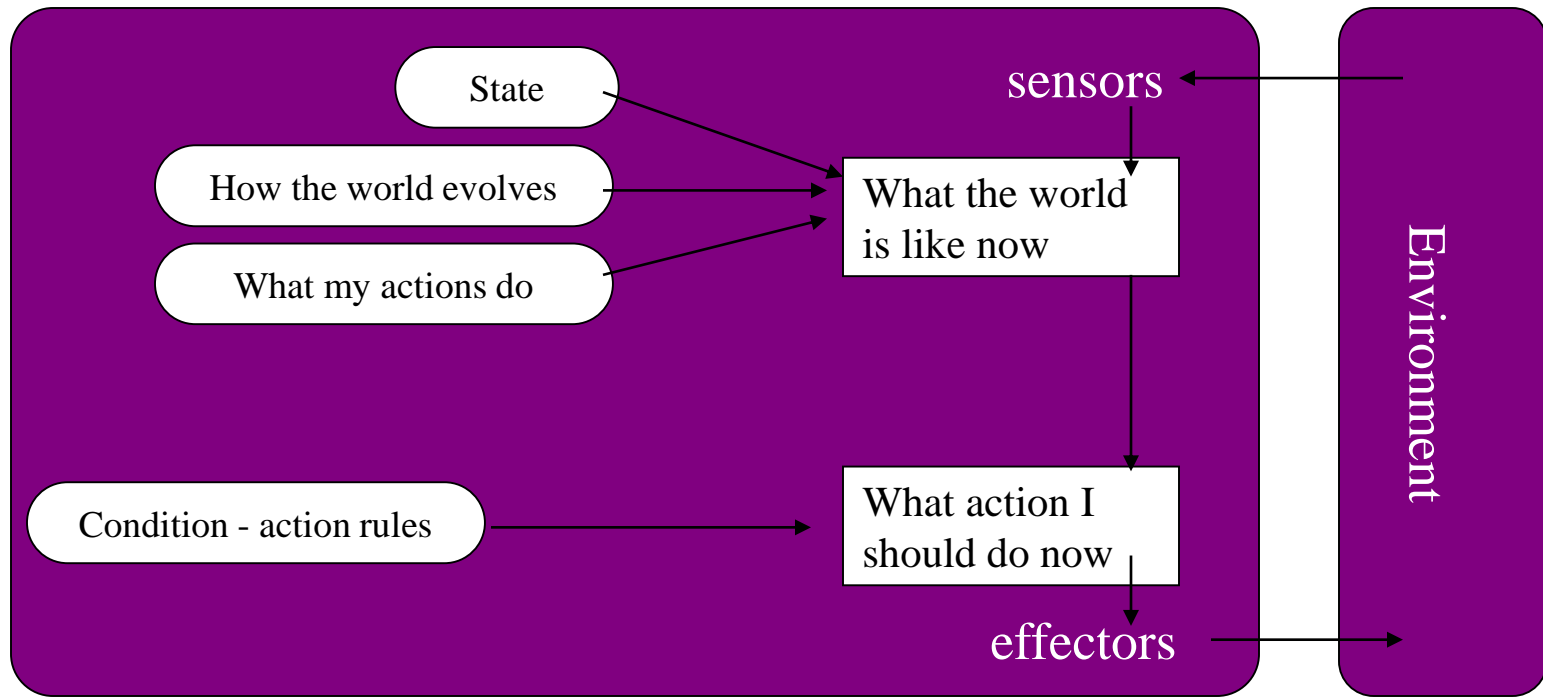
Needed for completeness

Action: SNAP or AVOID or NOOP

II. Model-Based Reflex Agents

- Maintains internal state that keeps track of aspects of the environment that cannot be currently observed
- For the world that is partially observable
 - **If the car is a recent model** - there is a centrally mounted brake light. With older models, there is no centrally mounted, so what if the agent gets confused?
Is it a parking light? Is it a brake light? Is it a turn signal light?
 - ✓ The agent has to keep track of an internal state
 - That depends on the percept history
 - Reflecting some of the unobserved aspects
 - Example, driving a car and changing lane
- Requiring two types of knowledge
 - ✓ How the world evolves independently of the agent
 - ✓ How the agent's actions affect the world
 - ✓ The agent is with memory

Figure: Structure of Model-Based reflex agent



function REFLEX-AGENT-WITH-STATE (*percept*) **returns** action

static: *state*, a description of the current world state

rules, a set of condition-action rules

state \leftarrow UPDATE-STATE (*state*, *percept*)

rule \leftarrow RULE-MATCH (*state*, *rules*)

action \leftarrow RULE-ACTION [*rule*]

state \leftarrow UPDATE-STATE (*state*, *action*)

return *action*

Example Table Agent With Internal State

IF

THEN

| | |
|---|---------------|
| Saw an object ahead, and turned right, and it's now clear ahead | Go straight |
| Saw an object Ahead, turned right, and object ahead again | Halt |
| See no objects ahead | Go straight |
| See an object ahead | Turn randomly |

■ The *Rules* are:-

- 1) If open(left) & open(right) and open(straight) then choose randomly between right and left
- 2) If wall(left) and open(right) and open(straight) then straight
- 3) If wall(right) and open(left) and open(straight) then straight
- 4) If wall(right) and open(left) and wall(straight) then left
- 5) If wall(left) and open(right) and wall(straight) then right
- 6) If wall(left) and door(right) and wall(straight) then open-door
- 7) If wall(right) and wall(left) and open(straight) then straight.
- 8) (Default) Move randomly

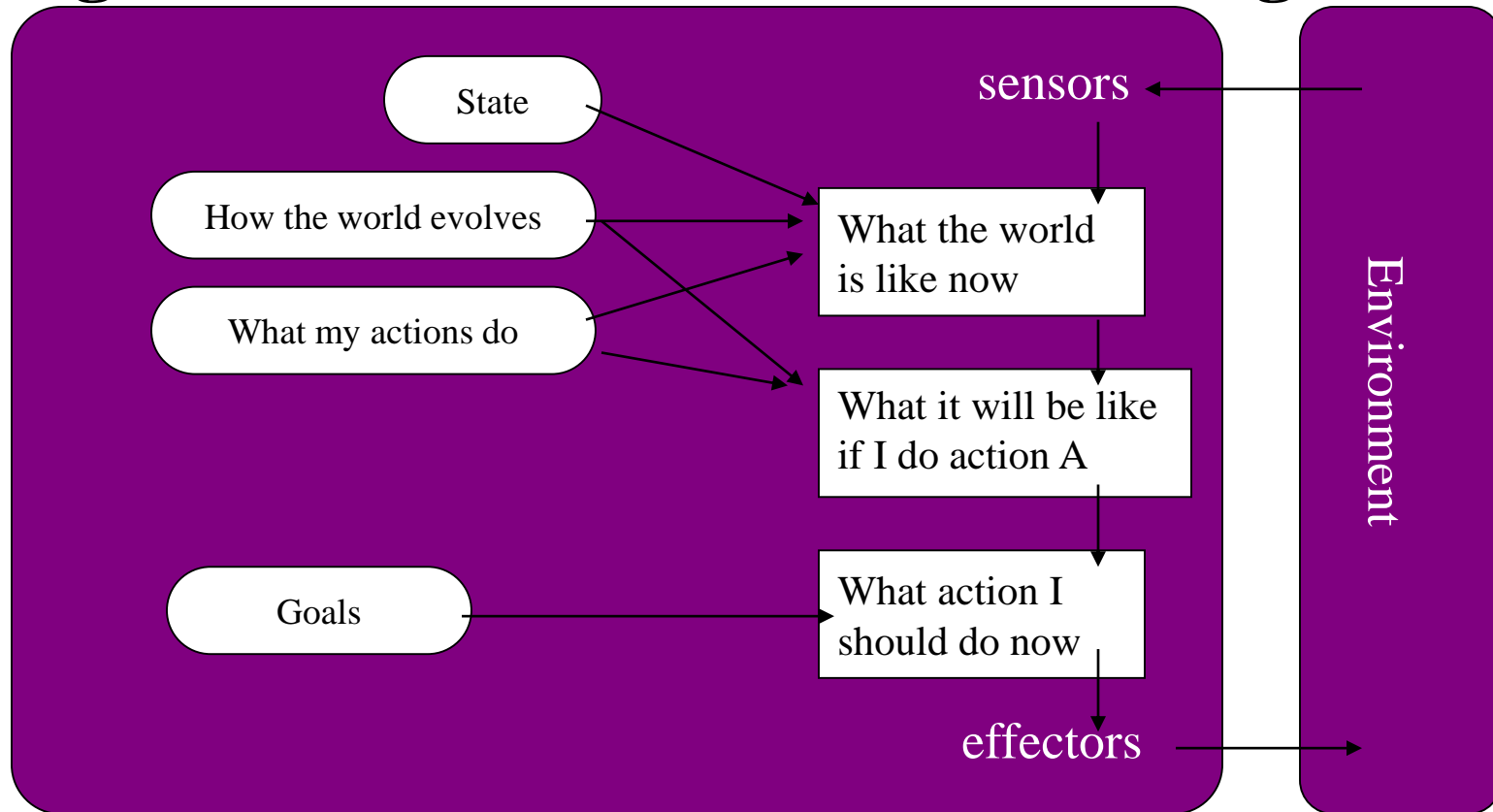
III. Goal-Based Agents

- The agent uses goal information to select between possible actions in the current state
- Current state of the environment is always not enough
- The goal is another issue to achieve
 - ✓ Judgment of rationality / correctness
- Actions chosen → goals, based on
 - ✓ The current state
 - ✓ The current percept
- E.g. At a road junction, the taxi can turn left, right or go straight.

■ Conclusion

- ✓ Goal-based agents are less efficient
- ✓ But more flexible
 - Agent ← Different goals ← different tasks
- ✓ Search and planning
 - Two other sub-fields in AI
 - To find out the action sequences to achieve its goal

Figure: Structure of a Goal-based agent

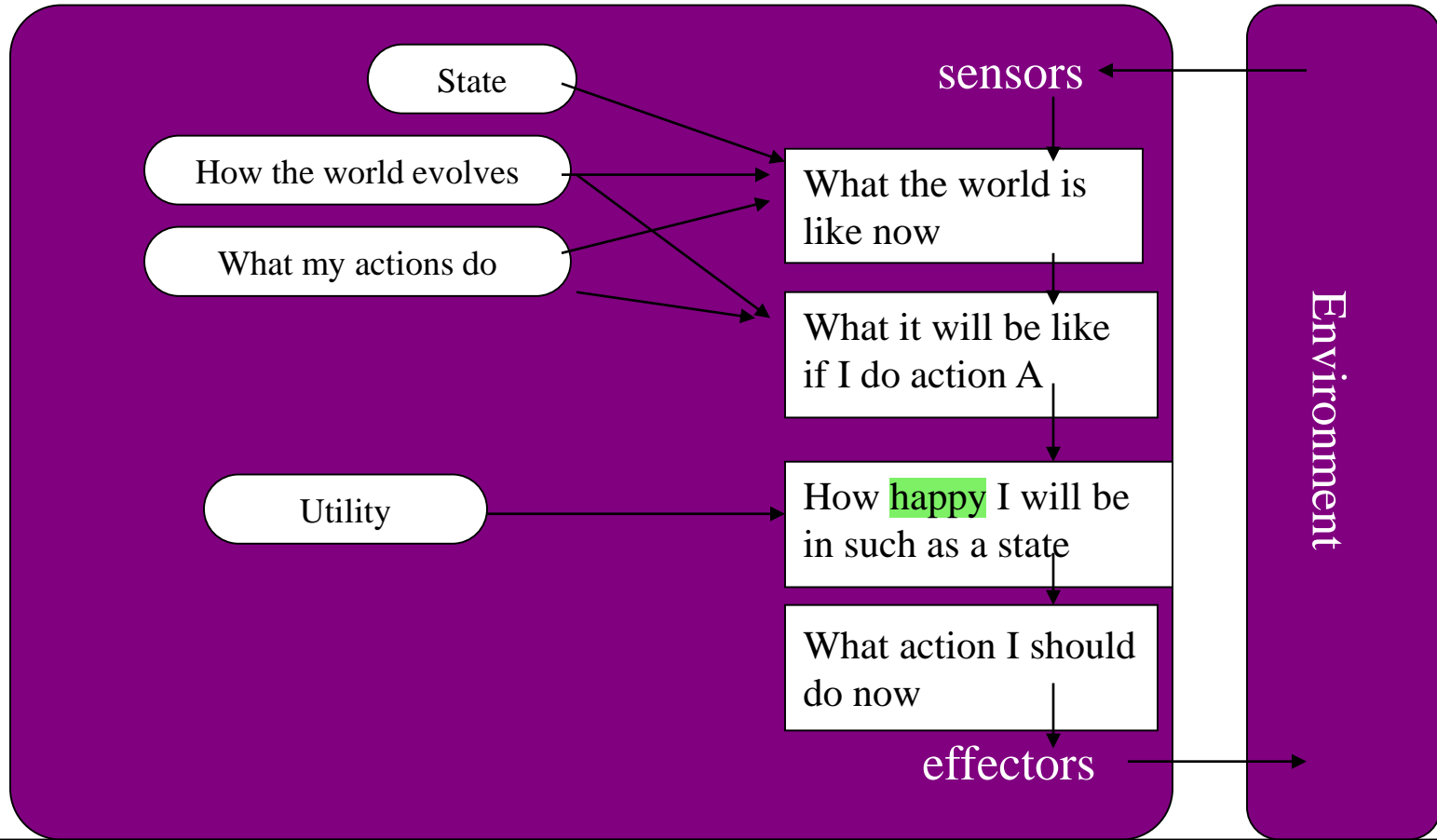


```
function GOAL_BASED_AGENT (percept) returns action
  state ← UPDATE-STATE (state, percept)
  action ← SELECT-ACTION [state, goal]
  state ← UPDATE-STATE (state, action)
  return action
```

IV. Utility-Based Agents

- The agent uses a utility function to evaluate the desirability of states that could result from each possible action
- Goals alone are not enough
 - ✓ To generate **high-quality** behavior
 - ✓ Example, meals in Canteen, good or not ?
- Many action sequences → the goals
 - ✓ Some are better and some worse
 - ✓ If goal means success,
 - ✓ Then **utility** means the degree of success (how successful it is).
- E.g. route recommendation system
 - There are many action sequences that will get the taxi to its destination, thereby achieving the goal.
 - Some are quicker, safer, more reliable, or cheaper than others. We need to consider **Speed and safety**

Figure: Structure of a utility-based agent



function UTILITY_BASED_AGENT (*percept*) **returns** action

state \leftarrow UPDATE-STATE (*state*, *percept*)

action \leftarrow SELECT-OPTIMAL_ACTION [*state*, *goal*]

state \leftarrow UPDATE-STATE (*state*, *action*)

return *action*

- It is said state A has higher utility
 - ✓ If state A is more preferred than others
- Utility is therefore a function
 - ✓ That maps a state onto a real number
 - ✓ The degree of success

- Utility has several advantages:
 - ✓ When there are conflicting goals,
 - Only some of the goals but not all can be achieved
 - utility describes the appropriate trade-off
 - ✓ When there are several goals
 - None of them are achieved **certainly**
 - Utility provides a way for the decision-making

V. Learning Agents

- To build learning machines and then to teach them.
- After an agent is programmed, can it work immediately?
 - ✓ No, it still need teaching
- In AI,
 - ✓ Once an agent is done
 - ✓ We teach it by giving it a set of examples
 - ✓ Test it by using another set of examples
- We then say the agent learns
 - ✓ A learning agent
- learning agents are able to perform tasks ,
analyze performance and look for new
ways to improve on those tasks

- Four conceptual components

- a) **Learning Element**

- ✓ Which is responsible for making improvement

- b) **Performance Element**

- ✓ Which is responsible for selecting external actions

- c) **Critic**

- ✓ The learning element uses feedback from the critic on how the agent is doing and determines how the performance element should be modified to do better in the future.

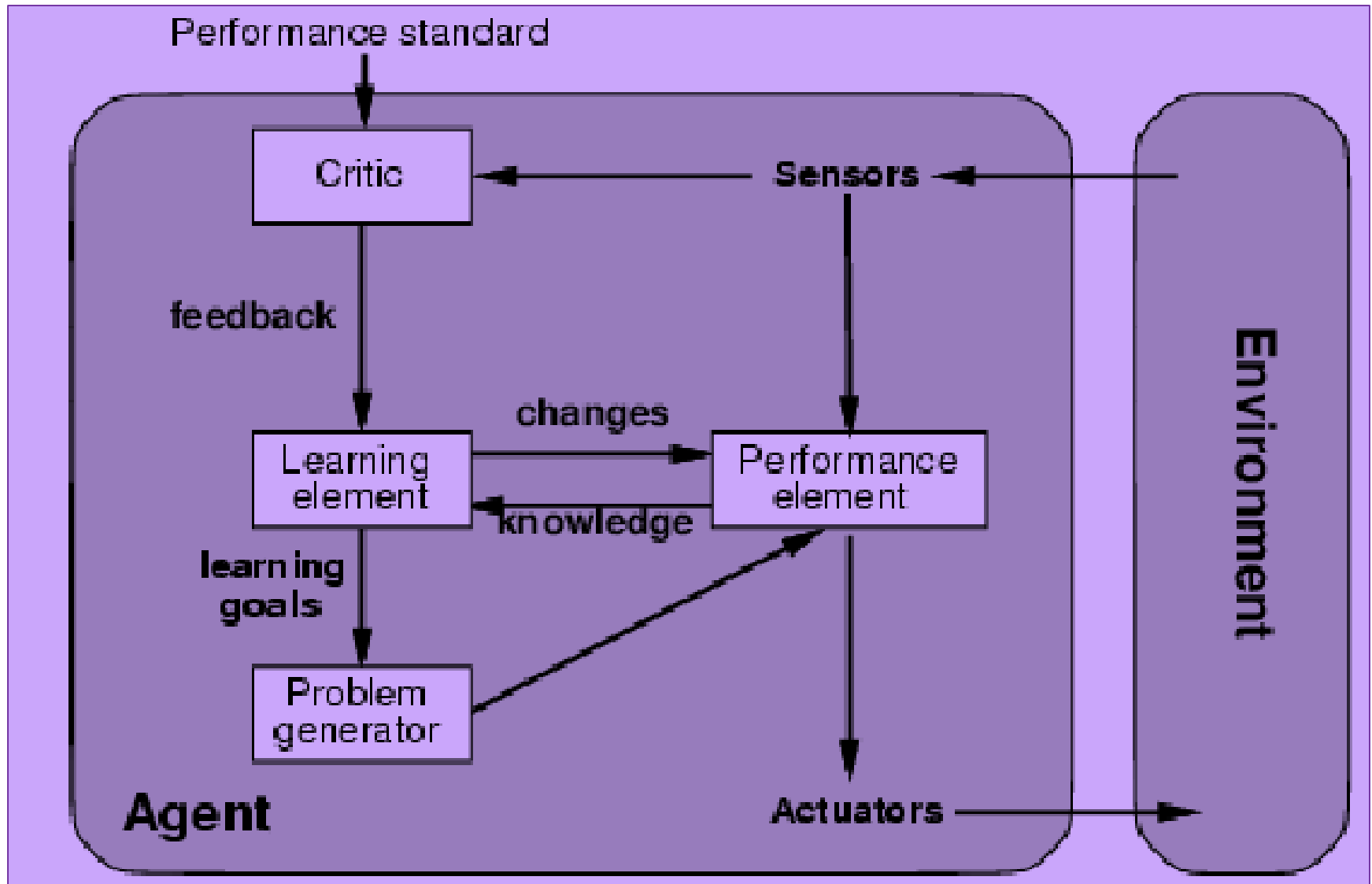
- ✓ Tells the Learning element how well the agent is doing with respect to fixed performance standard.

- Feedback from user or examples, good or not?

- d) **Problem Generator**

- ✓ Responsible for suggesting actions that will lead to new and informative experiences

Figure: Structure of a Learning agent



Summery hierarchy in types of agents

- **Reflex agents:**
 - ✓ These agents function in a current state, ignoring past history.
 - ✓ Responses are based on the event-condition-action rule ([ECA rule](#)) where a user initiates an event and the agent refers to a list of pre-set rules and pre-programmed outcomes.
- **Model-based agents:**
 - ✓ These agents choose an action in the same way as a reflex agent, but they have a more comprehensive view of the environment.
 - ✓ A model of the world is programmed into the internal system that incorporates the agent's history.
- **Goal-based agents:**
 - ✓ These agents expand upon the information model-based agents store by also including goal information, or information about desirable situations.

- **Utility-based agents:**

- ✓ These agents are similar to goal-based agents but **provide an extra utility measurement** which rates each possible scenario on its desired result and **chooses the action that maximizes the outcome.**
- ✓ **Rating criteria** examples could be the probability of success or the resources required.

- **Learning agents:**

- ✓ These agents have the ability to gradually improve and become more knowledgeable about an environment **over time** through an additional **learning element.**
- ✓ The learning element will use feedback to determine how performance elements should be changed to improve gradually.