

# Introduction to Computational Intelligence

## Lecture 2

# Outline

- Fundamental Definitions
  - Computational Intelligence Paradigms
    - ✓ ANN
    - ✓ Fuzzy systems
    - ✓ Evolutionary computation
    - ✓ Genetic algorithm
- Example Applications
- Summary
- Pop quiz

# Basic definitions Cont.

- **Evolutionary Computation (EC):**

- **Source of inspiration** - biological evolution and natural selection.

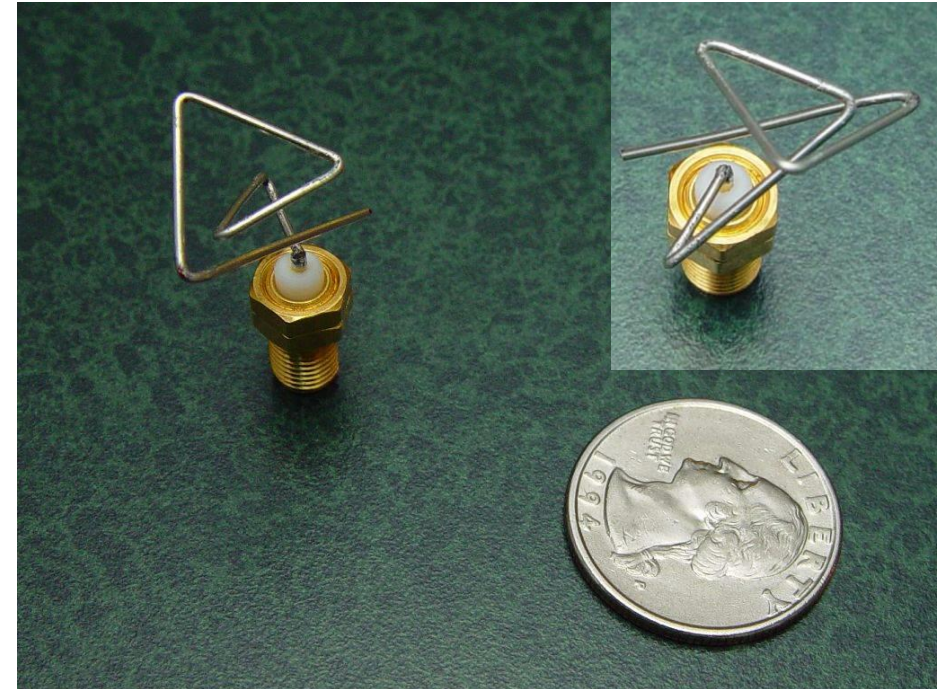
- **Model** - solves optimization problems by generating, evaluating and modifying a population of possible solutions.

- **Extensions** –

- ✓ Genetic algorithms (GA), evolutionary programming (EP),
    - ✓ Evolution strategies, genetic programming, swarm intelligence and optimization,
    - ✓ Differential evolution, evolvable hardware,
    - ✓ Multi-objective optimization
    - ✓ etc.

# Basic definitions Cont.

- **Example application of EC** –
  - NASA's Evolutionary Antenna ([Destination NASA - Evolutionary Antenna Synthesis - YouTube](#))
  - ✓ The complex shape of the antennas are designed by an EC-based design application to get the best radiation pattern.

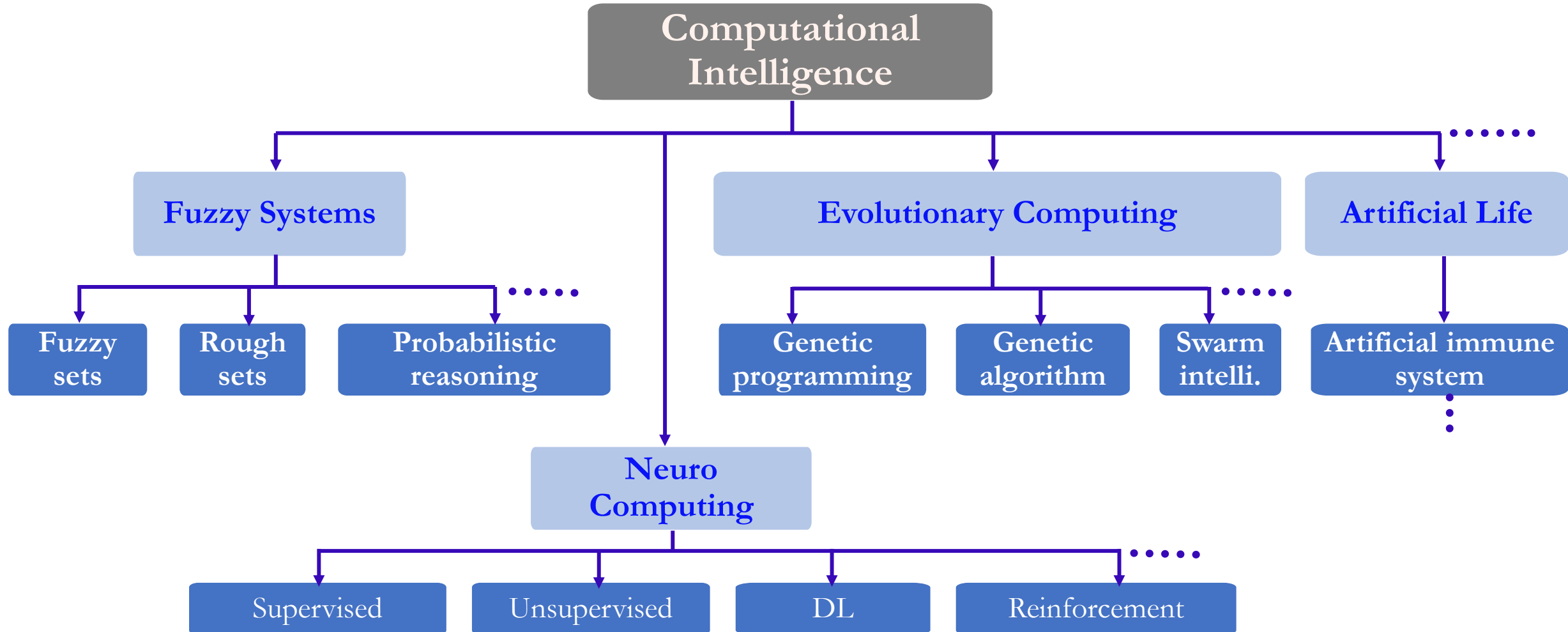


Source: <https://www.nasa.gov/>

# Basic definitions Cont.

- **Genetic algorithms (GA):**
- **Model** - incorporates natural evolution mechanisms, including crossover, mutation, and survival of the fittest.
- **EC vs GA:** Evolutionary computing algorithms are like genetic algorithms, but do not incorporate crossover.
- **Application:** They are more often used for optimization (search for the best), but also are used for classification.
  - Genetic algorithm tunes up public speakers ([Genetic algorithm tunes up public speakers | New Scientist](#))

# Computational Intelligence Paradigms



# CI Example Applications

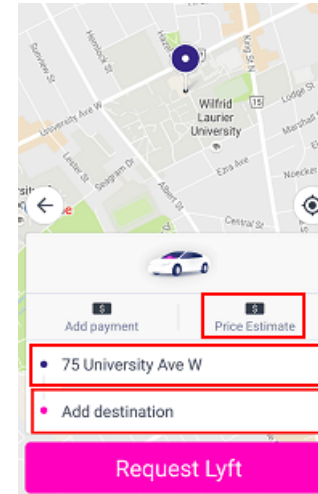
- **Google's AI-Powered Predictions:**

- Using anonymized [location data from smartphones](#)
- Google Maps can analyze the **speed of movement** of traffic at any given time.



- **Ridesharing Apps Like Uber and Lyft:**

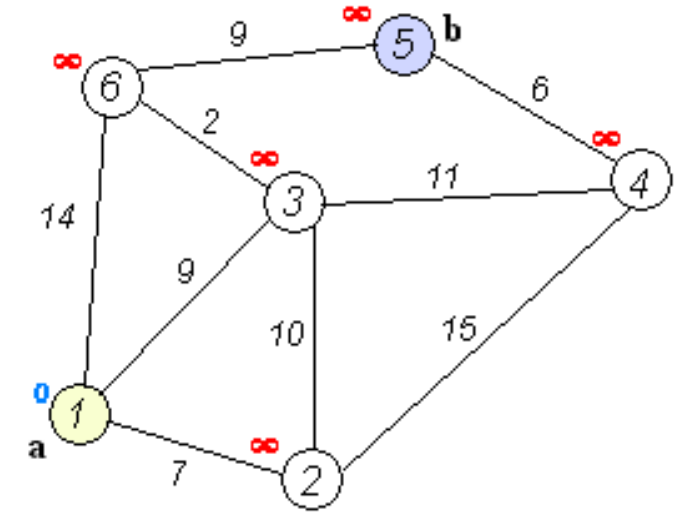
- How do they determine the price of your ride?
- How do they **minimize the wait time** once you hail a car?
- How do these services **optimally match** you with other passengers to minimize detours?
- The answer to all these questions is CI and ML



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- **Plagiarism Checkers:**

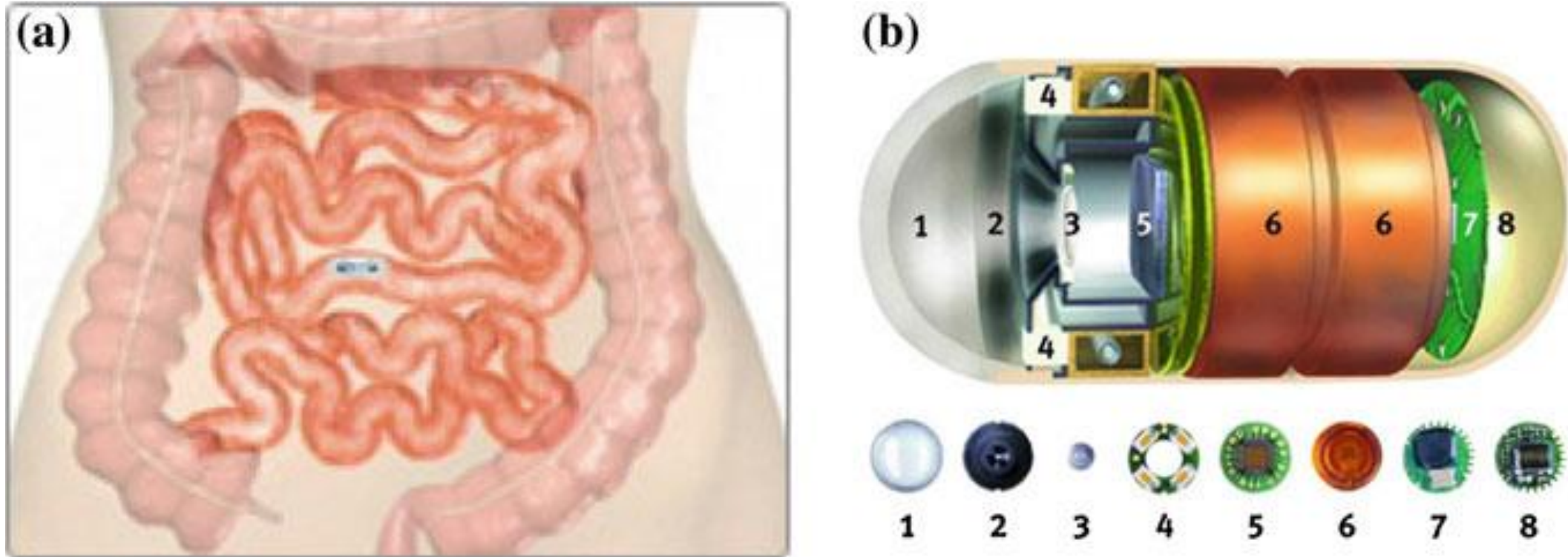
- Many high school and college students are familiar with services like *Turnitin*, a popular tool used by instructors to analyze students' writing for plagiarism.
- While *Turnitin* doesn't reveal precisely how it detects plagiarism, research demonstrates how CI can be used to develop a plagiarism detector.



<https://www.techemergence.com/everyday-examples-of-ai/> Image: Dijkstra's algorithm (Motherboard)



# CI Example Applications Cont.



**Figure:** Wireless capsule endoscopy: (a) The patient swallows the capsule and capsule travels through the tubular intestinal path. The imaging is done through a circular lens and at periodic intervals. (b) Pillcam Capsule parts:  
1 - Optical dome, 2 - Lens holder, 3 - Lens, 4 - Illuminating LEDs,  
5 - CMOS imager, 6 - Battery, 7 - ASIC transmitter, 8 - Antenna

Automatic Image Segmentation for Video Capsule Endoscopy: V.B. Surya Prasath and R. Delhibabu



# Summary

- The basic concept of CI was developed more than 45 years ago, but it took almost the last two decades for their potential to be recognized by a larger audience.
- Computational Intelligence-related techniques play an important role in state-of-the-art components and novel devices and services in science and engineering.
- The application of a unimodal CI technique will often be insufficient on its own to provide solutions to all the practical issues.
- Hybrid systems involving combinations of neural computation, fuzzy logic, and evolutionary algorithms, as well as traditional techniques, are a more promising approach for improving the performance of robot controllers.

Note: Go to course page → Quizzes and answer the following questions.

## Review Questions

1. Define computational intelligence.
2. Mention the different paradigms of computational intelligence.

# Intelligent Agent

## Lecture 2

# Agent

- An agent is anything that can be viewed as **perceiving** its **environment** through **sensors** and **acting** upon that environment through actuators



Can you suggest an example?

- **Human agent**

- eyes, ears, and other organs for sensors – hands, legs, mouth, and other body parts for actuators

- **Robotic agent**

- cameras and laser range finders for sensors – various motors for actuators

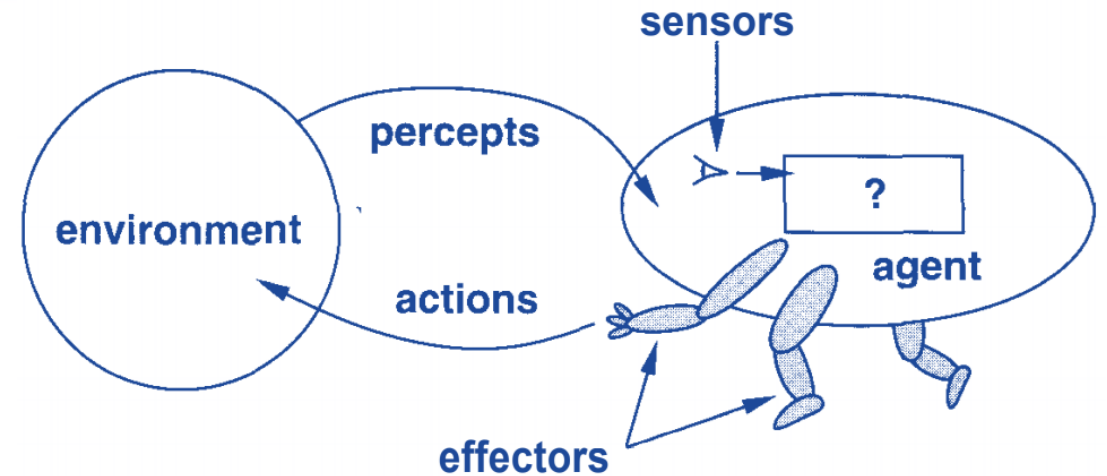


Figure: Agents interact with environments through sensors and effectors [2]



# Examples of Agents

- Intelligent buildings

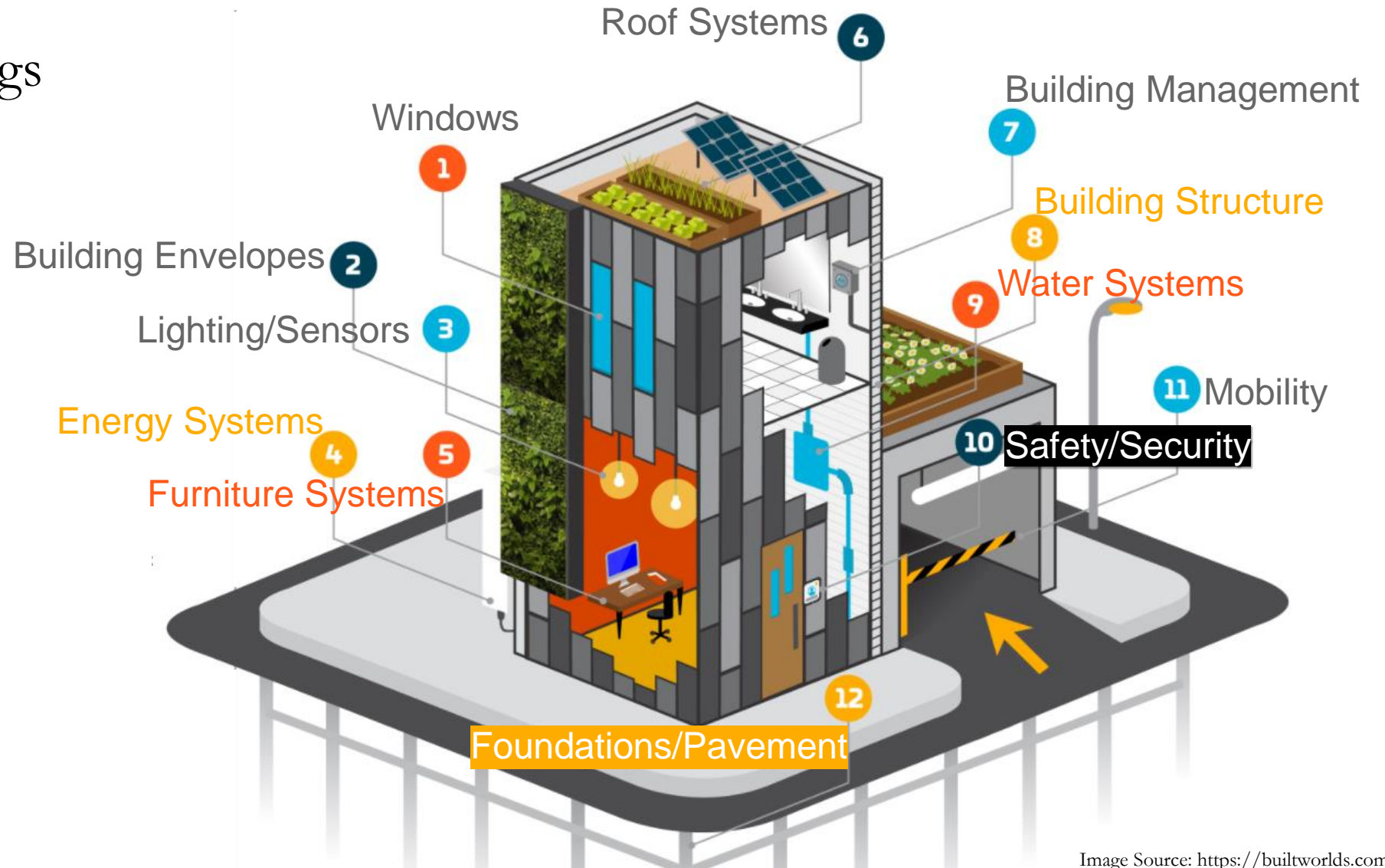


Image Source: <https://builtworlds.com/>



# Examples of Agents

- Robots
- Autonomous spacecraft

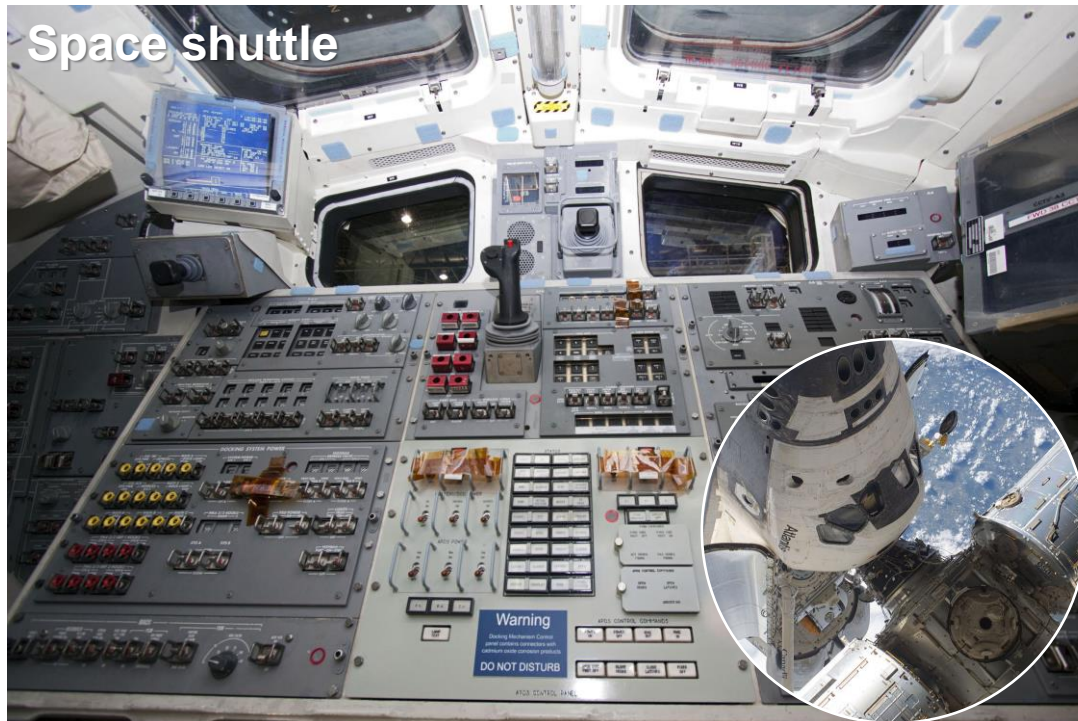


Image Sources: <https://www.technologyreview.com/>, <https://www.electronicweekly.com/>, Wikipedia.com



# Examples of Agents

- Web agents

- Powerful techniques to **process information** intelligently and **offer** features based on **patterns** and **relationships** in the data.
- Using the internet as a platform that not only gathers data at an ever-increasing pace but also systematically **transforms the raw data into actionable information**.

- E.g.,

- ✓ Netflix
- ✓ Amazon
- ✓ Google Ad Sense

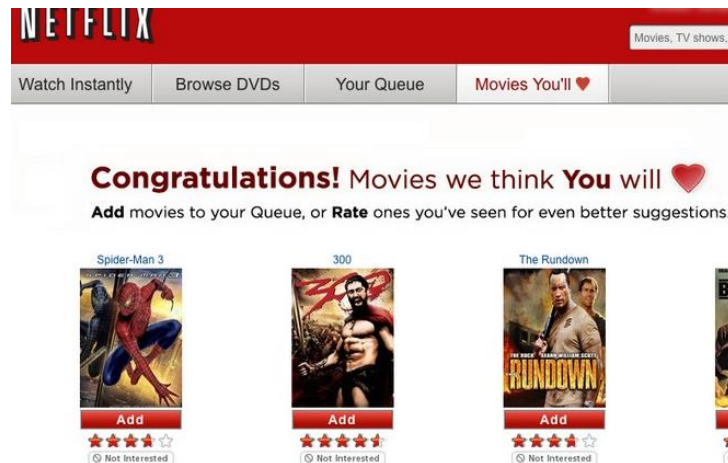


Image Sources: <https://blog.watsi.org/>, <http://www.webdesignerdepot.com/>



# Intelligent Agent's Objective

- **Optimize energy** and **increase utility**
  - Example objectives
    - ✓ Set a smart thermostat to turn on the A/C when building temps rise above 72°F
    - ✓ Use motion sensors to trigger lights in less-trafficked hallways to save energy
    - ✓ Install an access control system to improve building security and restrict access
    - ✓ Manage connected facilities through a Building Information Modeling (BIM) system

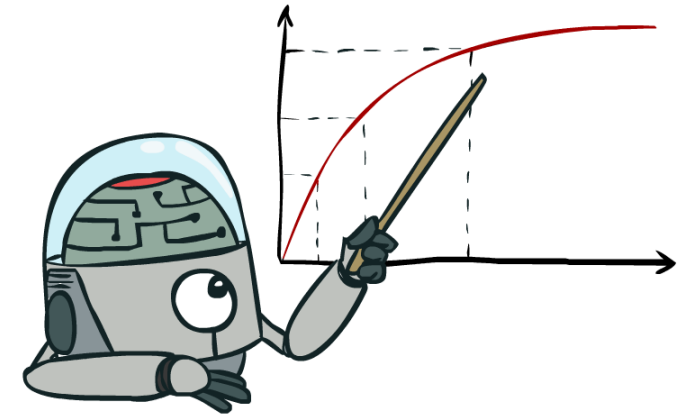


Image source: Anca Dragan's slides ([CS 188: Introduction to Artificial Intelligence, Fall 2020 \(berkeley.edu\)](#))

# Problem Solving Agents

- Intelligent agents are supposed to act in such a way that the environment goes through a sequence of states that **maximizes the performance measure**.
- This specification is **difficult to translate** into a successful agent design.
- Let's simplify it by adopting a **goal and aim to satisfy it**.

# Problem Solving Agents Cont.

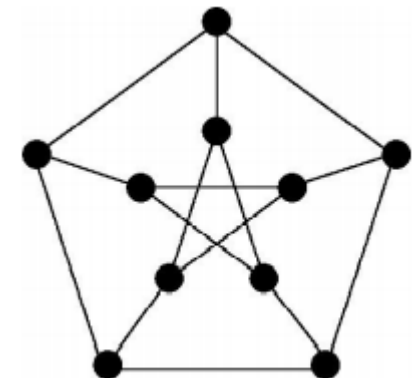
- **Goals** help organize behavior by **limiting objectives**.
- A **goal** to be a set of world **states** – exactly those states in which the goal is satisfied.
- **Assumption:** environments that are known, observable, discrete and deterministic (i.e., each action has exactly one corresponding outcome).
  - action  $A \rightarrow$  outcome  $R$
- The process of **looking for a sequence of actions** that **reaches** the **goal** is called **search**.
- A search algorithm takes a **problem space as input and returns a solution** in the form of an **action sequence**.

# Formulating Problems and Solutions

- A problem can be defined formally by (5) components:
  1. The initial state from which the agent starts
  2. Operator : a description of possible actions available to the agent:  
 $ACTIONS(s)$
  3. A description of what each action does, i.e., the **transition model**, specified by a function  $RESULT(s, a) = a'$ .

## State space:

- The **initial state**, **actions** and **transition model** implicitly defined the state space of the problem – the set of all states reachable from the initial state by any sequence of actions.
- It forms a **directed network** or graph in which the nodes are states and the edges between nodes are actions.
  - A **path** in the state space is a sequence of states connected by a sequence of actions.



# Formulating Problems and Solutions Cont.

4. **Goal test:** it determines whether a given state is a **goal state**.
  - Frequently the **goal test is intuitive** (e.g., check if we arrived at the destination) – but note that it is also sometimes specified by an abstract property (e.g., “check mate”).
  
5. **Path cost function:** It assigns a numeric cost to each path.
  - The agent chooses a cost function that reflects its own performance measure.
  - Commonly (but not always), the cost of a path is additive in terms of the individual actions along a path.
  - Denote the step cost to take action ‘ $a$ ’ in state  $s$ , arriving in  $s'$  as:  $c(s, a, s')$ .