

# Introduction to Artificial Intelligence

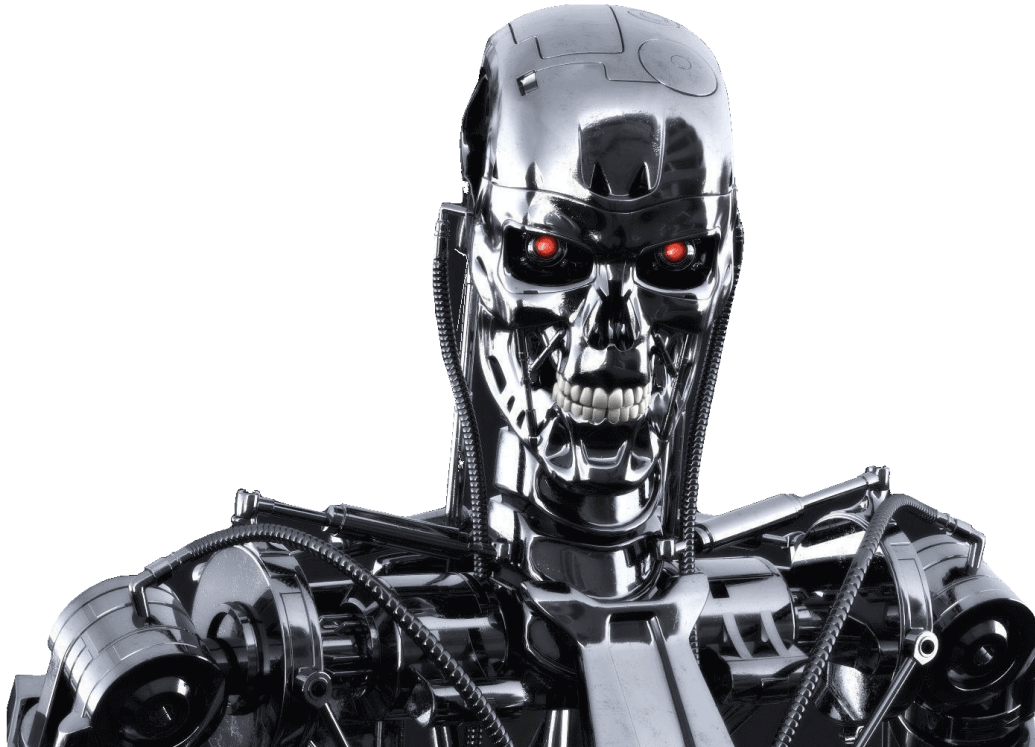
Lecture 1: Foundations

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# Today

- Course outline
- Introduction to Artificial Intelligence
- Intelligent agents

# Introduction



"With artificial intelligence we are summoning the demon" -- Elon Musk



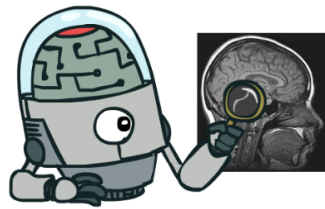
"We're really closer to a smart washing machine than Terminator" -- Fei-Fei Li,  
Director of Stanford AI Lab.

# What is AI?

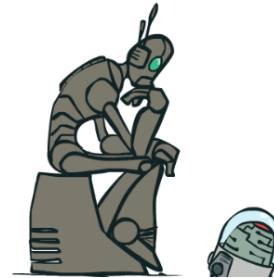
# What is AI?

Artificial intelligence is the science of making machines or programs that:

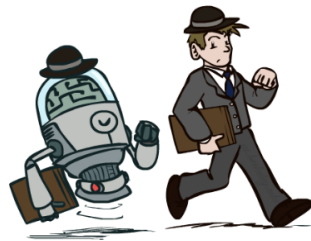
*Think like people*



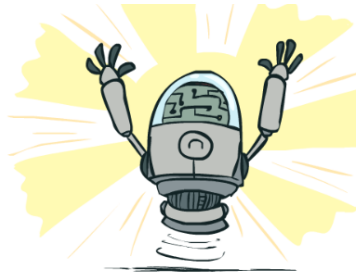
*Think rationally*



*Act like people*



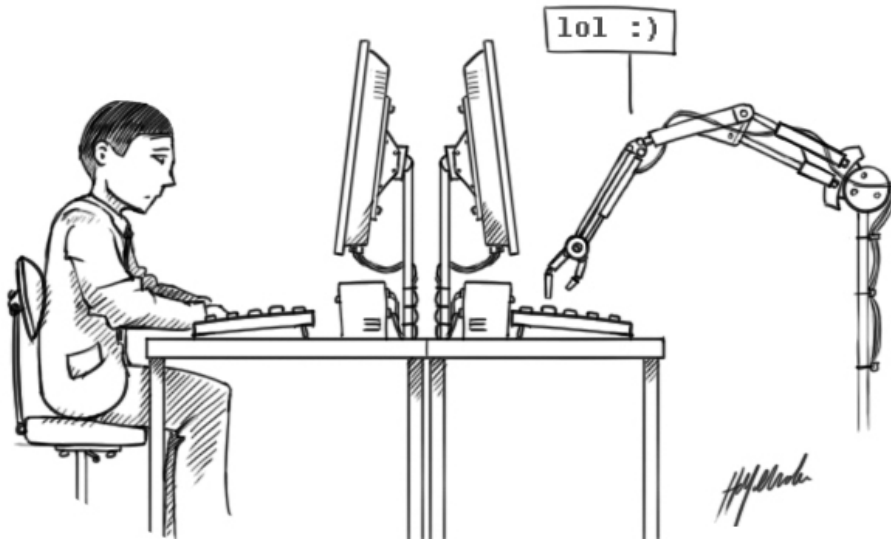
*Act rationally*



# Acting humanly

## The Turing test

A computer passes the **Turing test** (aka the Imitation Game) if a human operator, after posing some written questions, cannot tell whether the written responses come from a person or from a computer.



*Can machines think?*  
(Alan Turing, 1950)



An agent would not pass the Turing test without the following requirements:

- natural language processing
- knowledge representation
- automated reasoning
- machine learning
- computer vision (total Turing test)
- robotics (total Turing test)

Despite being proposed almost 70 years ago, the Turing test is still relevant today.

## Limitations of the Turing test

The Turing test tends to focus on **human-like errors**, **linguistic tricks**, etc.

However, it seems more important to study the **principles** underlying intelligence than to replicate an exemplar.



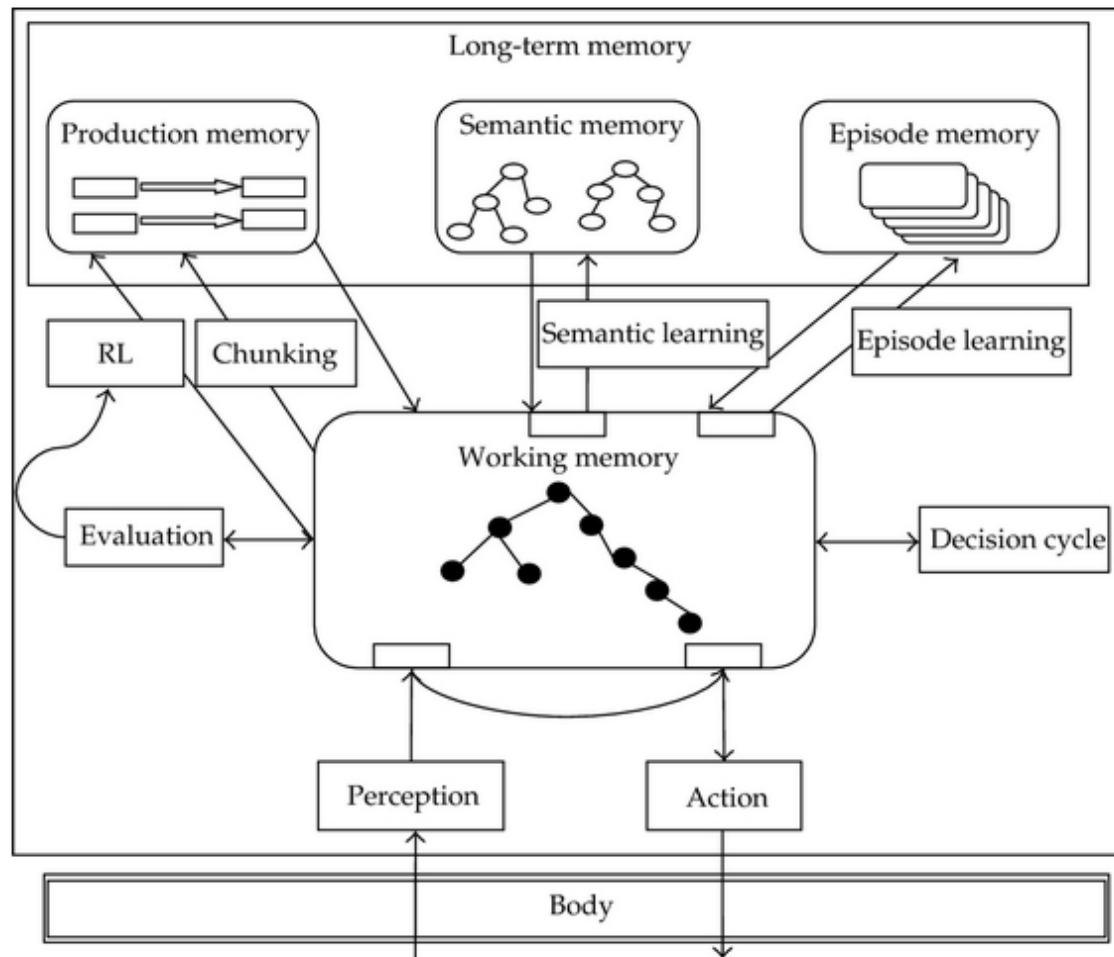
*Aeronautics is not defined as the field of making machines that fly so exactly like pigeons that they can fool even other pigeons.*

# Thinking humanly

## Cognitive science

Study of the **human mind** and its processes.

- Grew out of human psychology.
- It includes language, problem-solving, decision-making and perception.
- The goal of cognitive science is to form a theory about the structure of the mind, summarized as a comprehensive **computer model**.
- A **cognitive architecture** usually follows human-like reasoning and can be used to produce testable predictions (time of delays during problem solving, kinds of mistakes, learning rates, etc).

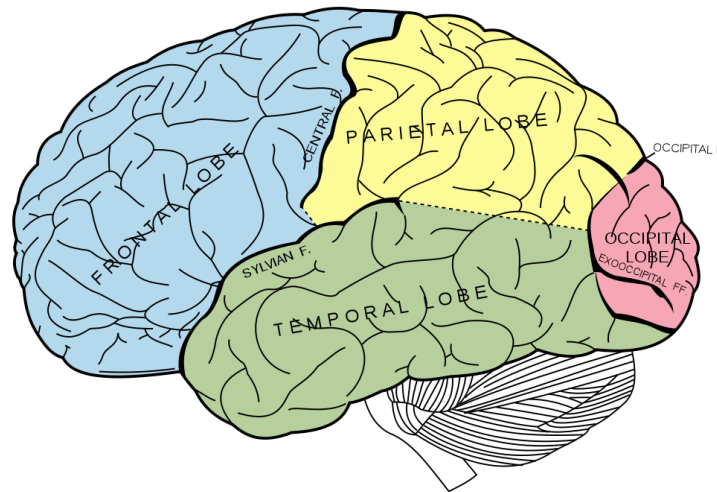


*The modern SOAR cognitive architecture.*

# Neuroscience

Study of the anatomy and physiology of neural tissue.

- Grew out of clinical neurology and neurobiology.
- Neurobiology is concerned with the the anatomy and physiology of the brain, from major structures down to neurons and molecules.
- Neuroscience adds to that the study of **how the brain works**, mechanistically, functionally, and systematically to produce observable behavior.



## Limitations of cognition and neuroscience for AI

- In linguistics, the argument of **poverty of the stimulus** states that children do not receive sufficient input to generalize grammatical rules through linguistic input alone.
  - A baby hears too few sentences to deduce the grammar of English before he speaks correctly.
- (Controversial) Therefore, humans must be **biologically pre-wired** with **innate knowledge** for representing language.



*How do we know what we know?  
(Noam Chomsky, 1980)*

It may not be possible to implement a fully functioning computer model of the human mind without background knowledge of some sort. This is a huge technical **obstacle**, as accessing this knowledge would require reverse-engineering the brain.

# Thinking rationally

## The logical approach

- The rational thinking approach is concerned with the study of **irrefutable reasoning processes**. It ensures that all actions performed by a computer are formally **provable** from inputs and prior knowledge.
- The "laws of thought" were supposed to govern the operation of the mind. Their study initiated the field of **logic** and the **logician tradition** of AI (1960-1990).

## The Zebra puzzle

- There are five houses.
- The English man lives in the red house.
- The Swede has a dog.
- The Dane drinks tea.
- The green house is immediately to the left of the white house.
- They drink coffee in the green house.
- The man who smokes Pall Mall has birds.
- In the yellow house they smoke Dunhill.
- In the middle house they drink milk.
- The Norwegian lives in the first house.
- The man who smokes Blend lives in the house next to the house with cats.
- In a house next to the house where they have a horse, they smoke Dunhill.
- The man who smokes Blue Master drinks beer.
- The German smokes Prince.
- The Norwegian lives next to the blue house.
- They drink water in a house next to the house where they smoke Blend.



Who owns the zebra?

```
select([A|As],S):- select(A,S,S1),select(As,S1).
select([],_).
```

```
next_to(A,B,C):- left_of(A,B,C) ; left_of(B,A,C).
left_of(A,B,C):- append(_,[A,B|_],C).
```

```
zebra(Owns, HS):- % color,nation,pet,drink,smokes
    HS = [ h(_,norwegian,_,_,_), _, h(_,_,_,milk,_), _, _],
    select( [ h(red,englishman,_,_,_), h(_,swede,dog,_,_),
              h(_,dane,_,tea,_), h(_,german,_,_,prince) ], HS),
    select( [ h(_,_,birds,_,pallmall), h(yellow,_,_,_,dunhill),
              h(_,_,_,beer,bluemaster) ], HS),
    left_of( h(green,_,_,coffee,_), h(white,_,_,_,_), HS),
    next_to( h(_,_,_,_,dunhill), h(_,_,horse,_,_), HS),
    next_to( h(_,_,_,_,blend), h(_,_,cats,_,_), HS),
    next_to( h(_,_,_,_,blend), h(_,_,_,water,_), HS),
    next_to( h(_,norwegian,_,_,_), h(blue,_,_,_,_), HS),
    member( h(_,Owns,zebra,_,_), HS).
```

```
:- ?- time(( zebra(Who, HS), maplist(writeln,HS), nl, write(Who), nl, nl,
        ; write('No more solutions.') )).
```

Output =

```
h(yellow,norwegian, cats,  water, dunhill)
h(blue,  dane,      horse, tea,  blend)
h(red,   englishman,birds, milk,  pallmall)
h(green, german,    zebra, coffee,prince)
h(white, swede,     dog,   beer,  bluemaster)
```

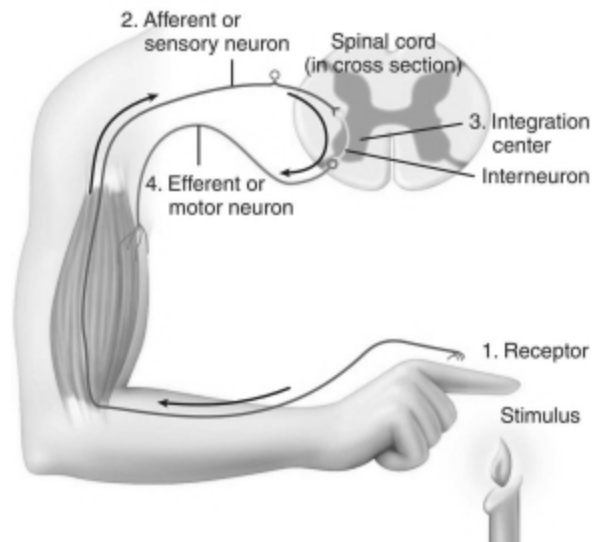
german

No more solutions.

% 5,959 inferences, 0.000 CPU in 0.060 seconds (0% CPU, Infinite Lips)

## Limitations of logical inference

- Representation of **informal** knowledge is difficult.
- Hard to define provable **plausible** reasoning.
- **Combinatorial explosion** (in time and space).
- Logical inference is only a part of intelligence. It does not cover everything:
  - e.g., might be no provably correct thing to do, but still something must be done;
  - e.g., reflex actions can be more successful than slower carefully deliberated ones.



*Pain withdrawal reflexes do not involve inference.*

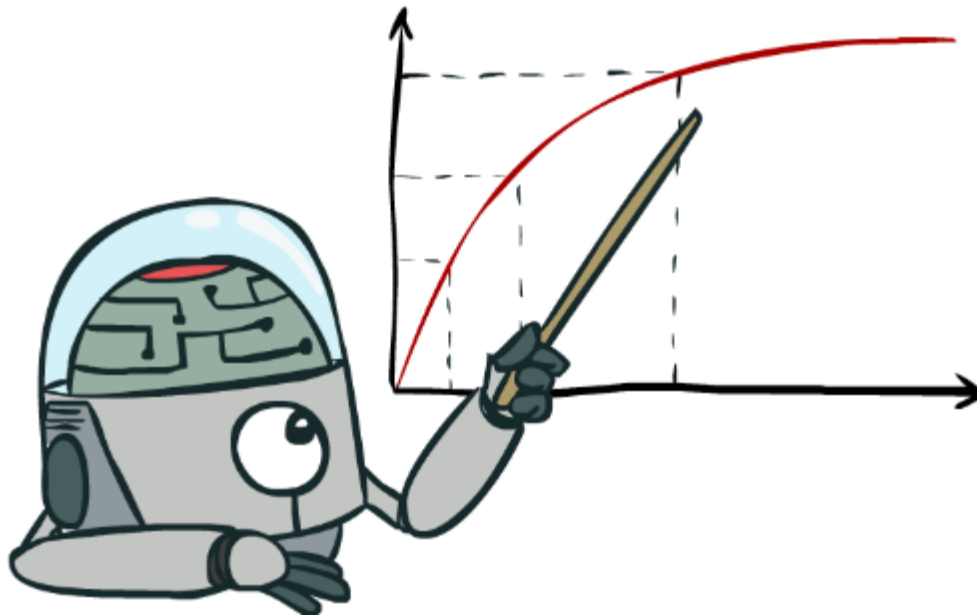
# Acting rationally

A **rational agent** acts so as to achieve the best (expected) outcome.

- Correct logical inference is just one of several possible mechanisms for achieving this goal.
- Perfect rationality cannot be achieved due to computational limitations! The amount of reasoning is adjusted according to available resources and importance of the result.
- The brain is good at making rational decisions but not perfect either.

Rationality only concerns **what** decisions are made (not the thought process behind them, human-like or not).

Goals are expressed in terms of the **performance** or **utility** of outcomes. Being rational means maximizing its expected performance. The standard of rationality is general and mathematically well defined.



In this course, Artificial intelligence = **Maximizing expected performance**

# AI prehistory

- **Philosophy:** logic, methods of reasoning, mind as physical system, foundations of learning, language, rationality.
- **Mathematics:** formal representation and proof, algorithms, computation, (un)decidability, (in)tractability, probability.
- **Psychology:** adaptation, phenomena of perception and motor control, psychophysics.
- **Economics:** formal theory of rational decisions.
- **Linguistics:** knowledge representation, grammar.
- **Neuroscience:** plastic physical substrate for mental activity.
- **Control theory:** homeostatic systems, stability, simple optimal agent designs.

# A short history of AI

## **1940-1950: Early days**

- 1943: McCulloch and Pitts: Boolean circuit model of the brain.
- 1950: Turing's "Computing machinery and intelligence".

## **1950-1970: Excitement and expectations**

- 1950s: Early AI programs, including Samuel's checkers program, Newell and Simon's Logic Theorist and Gelernter's Geometry Engine.
- 1956: Dartmouth meeting: "Artificial Intelligence" adopted.
- 1958: Rosenblatt invents the perceptron.
- 1965: Robinson's complete algorithm for logical reasoning.
- 1966-1974: AI discovers computational complexity.





## **1970-1990: Knowledge-based approaches**

- 1969: Neural network research almost disappears after Minsky and Papert's book.
- 1969-1979: Early development of knowledge-based systems.
- 1980-1988: Expert systems industrial boom.
- 1988-1993: Expert systems industry busts (AI winter).

## **1990-Present: Statistical approaches**

- 1985-1995: The return of neural networks.
- 1988-: Resurgence of probability, focus on uncertainty, general increase in technical depth.
- 1995-2010: New fade of neural networks.
- 1995-: Complete intelligent agents and learning systems.
- 2000-: Availability of very large datasets.
- 2010-: Availability of fast commodity hardware (GPUs).
- 2012-: Resurgence of neural networks with deep learning approaches.

# What can AI do at present?

- Translate spoken Chinese to spoken English, live?
- Answer multi choice questions, as good as an 8th grader?
- Converse with a person for an hour?
- Play decently at Chess? Go? Poker? Soccer?
- Buy groceries on the web? in a supermarket?
- Prove mathematical theorems?
- Drive a car safely on a parking lot? in New York?
- Perform a surgery?
- Identify skin cancer better than a dermatologist?
- Write a funny story?
- Paint like Vangogh? Compose music?
- Show common sense?



Playing Atari games



Beat the best human Go players



Beat teams of human players at real-time strategy games (Dota 2)



Speech translation and synthesis



Semantic segmentation





Generating image descriptions



Detecting skin cancer



Learning to walk



Folding laundry



Playing soccer



Driving a car



Learning to sort waste (Norman Marlier, ULiège, 2018)



Learning to sort waste (Norman Marlier, ULiège, 2018)



# What is missing?

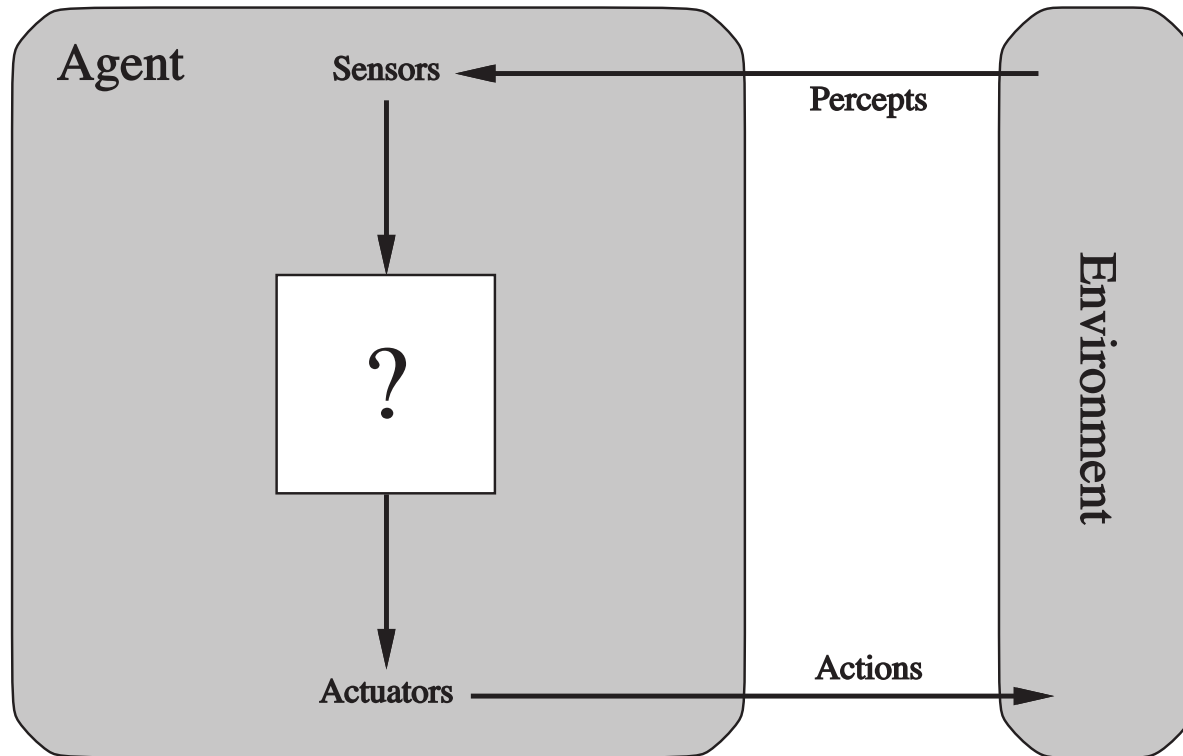
Intelligence is not just about **pattern recognition**, which is something most of these works are based on.

It is about **modeling the world**:

- explaining and understanding what we see;
- imagining things we could see but haven't yet;
- problem solving and planning actions to make these things real;
- building new models as we learn more about the world.

# Intelligent agents

# Agents and environments



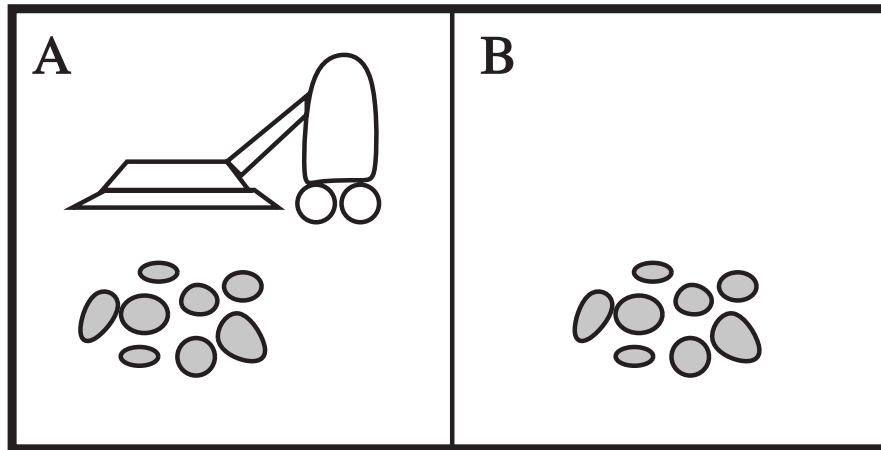
## Agents

- An **agent** is an entity that **perceives** its environment through sensors and take **actions** through actuators.
- The agent behavior is described by the **agent function**, or **policy**, that maps percept histories to actions:

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

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

## Vacuum-cleaner world



- Percepts: location and content, e.g.  $[A, \text{Dirty}]$
- Actions: **Left**, **Right**, **Suck**, **NoOp**

## A vacuum-cleaner agent

Partial tabulation of a simple vacuum-cleaner agent function:

Percept sequence	Action
[A, Clean]	Right
[A, Dirty]	Suck
[B, Clean]	Left
[A, Dirty]	Suck
[A, Clean], [A, Clean]	Right
[A, Clean], [A, Dirty]	Suck
(...)	(...)

## The optimal vacuum-cleaner?

What is the **right** agent function?

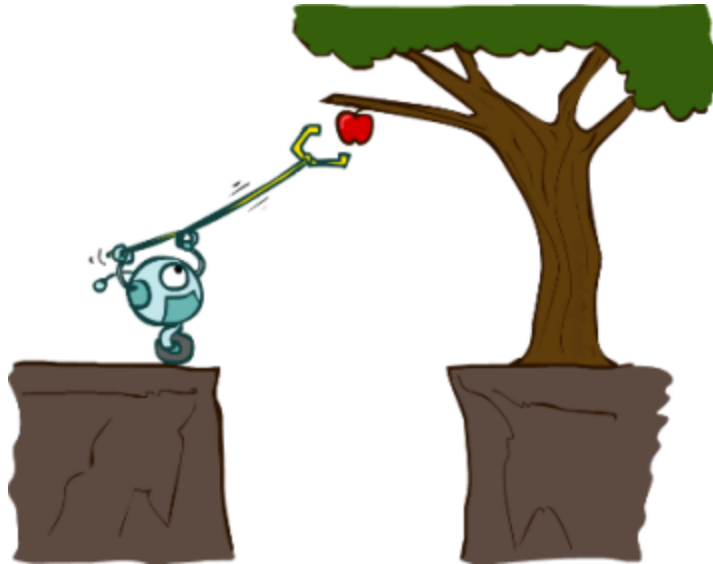
How to formulate the **goal** of the vacuum-cleaner agent?

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

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

# Rational agents

- Informally, a **rational agent** is an agent that does the "right thing".
- A **performance measure** evaluates a sequence of environment states caused by the agent's behavior.
- A rational agent is an agent that chooses whichever action that **maximizes** the **expected** value of the performance measure, given the percept sequence to date.





- Rationality  $\neq$  omniscience
  - percepts may not supply all relevant information.
- Rationality  $\neq$  clairvoyance
  - action outcomes may not be as expected.
- Hence, rational  $\neq$  successful.
- However, rationality leads to exploration, learning and autonomy.

# Performance, environment, actuators, sensors

The characteristics of the performance measure, environment, action space and percepts dictate techniques for selecting rational actions.

These characteristics are summarized as the **task environment**.

## Example 1: an autonomous car

- **performance measure**: safety, destination, legality, comfort, ...
- **environment**: streets, highways, traffic, pedestrians, weather, ...
- **actuators**: steering, accelerator, brake, horn, speaker, display, ...
- **sensors**: video, accelerometers, gauges, engine sensors, GPS, ...

## Example 2: an Internet shopping agent

- **performance measure**: price, quality, appropriateness, efficiency
- **environment**: current and future WWW sites, vendors, shippers
- **actuators**: display to user, follow URL, fill in form, ...
- **sensors**: web pages (text, graphics, scripts)

# Environment types

- Fully observable vs. partially observable
  - Whether the agent sensors give access to the complete state of the environment, at each point in time.
- Deterministic vs. stochastic
  - Whether the next state of the environment is completely determined by the current state and the action executed by the agent.
- Episodic vs. sequential
  - Whether the agent's experience is divided into atomic independent episodes.
- Static vs. dynamic
  - Whether the environment can change, or the performance measure can change with time.
- Discrete vs. continuous
  - Whether the state of the environment, the time, the percepts or the actions are continuous.
- Single agent vs. multi-agent
  - Whether the environment include several agents that may interact which each other.
- Known vs unknown
  - Reflects the agent's state of knowledge of the "law of physics" of the environment.

Are the following task environments fully observable? deterministic? episodic? static? discrete? single agents? Known?

- Crossword puzzle
- Chess, with a clock
- Poker
- Backgammon
- Taxi driving
- Medical diagnosis
- Image analysis
- Part-picking robot
- Refinery controller
- The real world

# Agent programs

The job of AI is to design an **agent program** that implements the agent function. This program will run on an **architecture**, that is a computing device with physical sensors and actuators.

$$\text{agent} = \text{program} + \text{architecture}$$

## Implementation

Agent programs can be designed and implemented in many ways:

- with tables
- with rules
- with search algorithms
- with learning algorithms

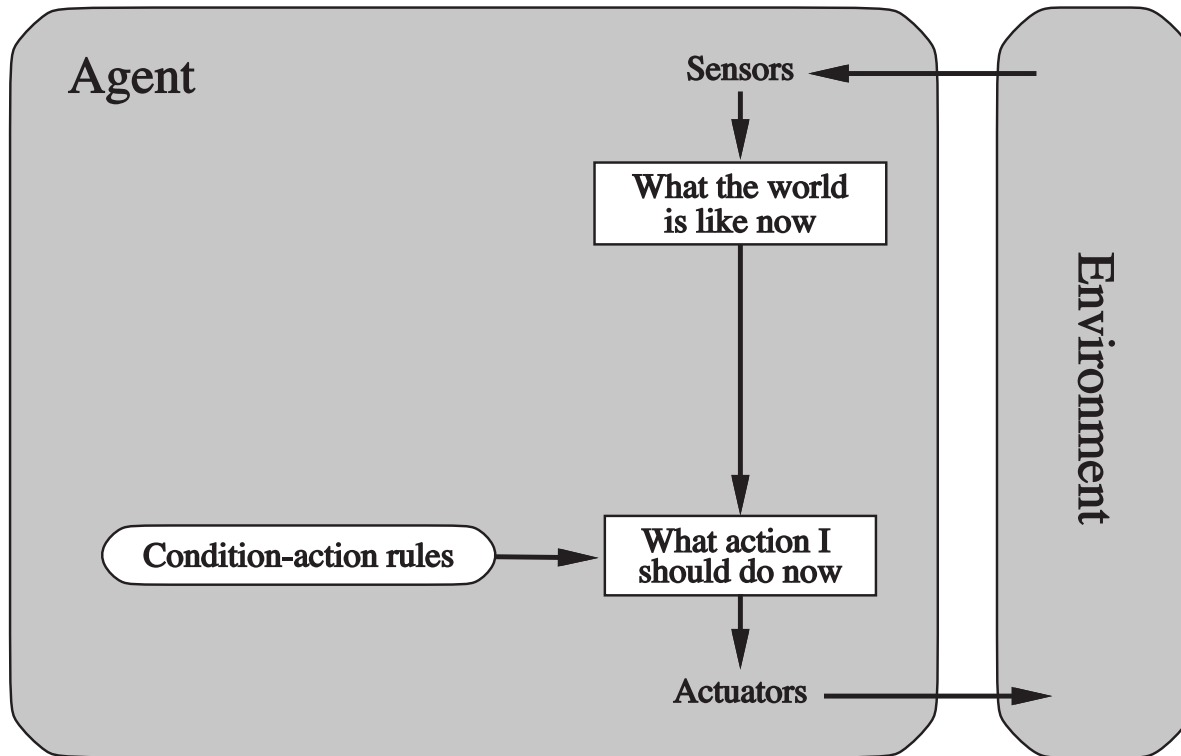
# Table-driven agents

A **table-driven agent** determines its next action with a lookup table that contains the appropriate action for every possible percept sequence.

## Issues

- **Design issue:** one needs to anticipate all sequence of percepts and how the agent should respond.
- **Technical issue:** the lookup table will contain  $\sum_{t=1}^T |\mathcal{P}|^t$  entries.
- Example (autonomous car): using a 30fps 640x480 RGB camera as sensor, this results in a table with over  $10^{250000000000}$  entries for an hour of driving.

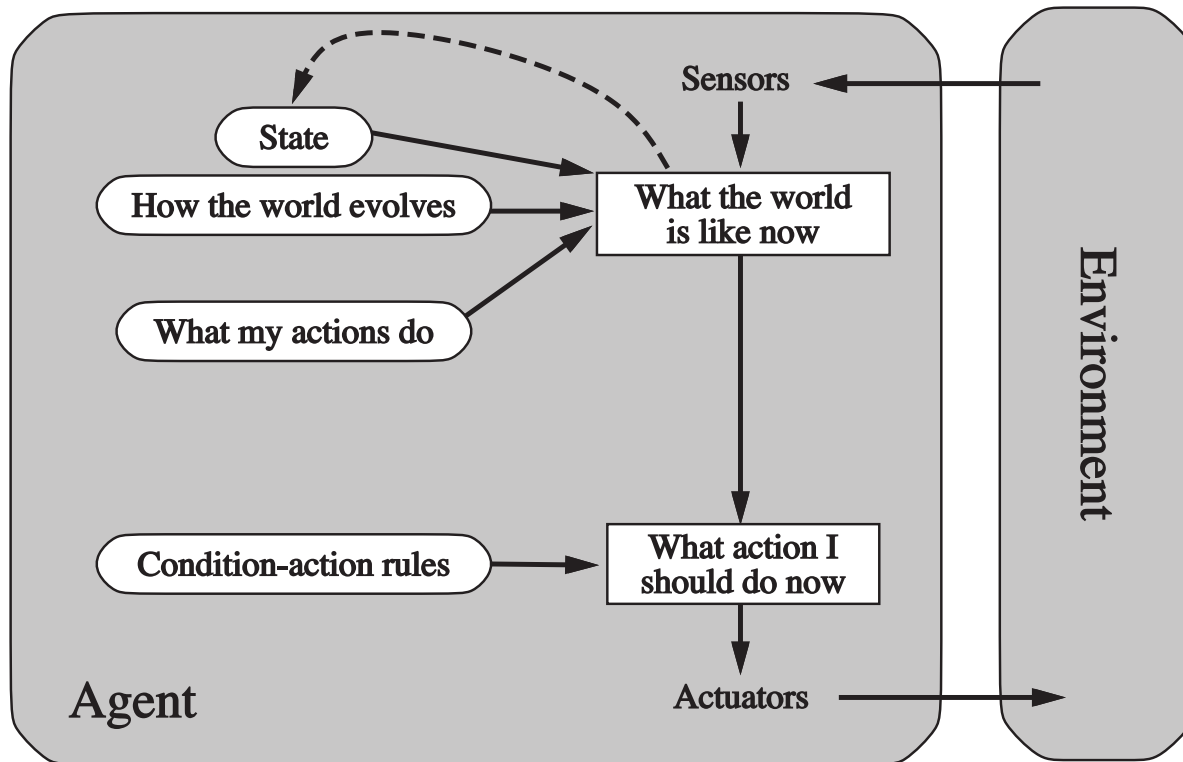
# Simple reflex agents





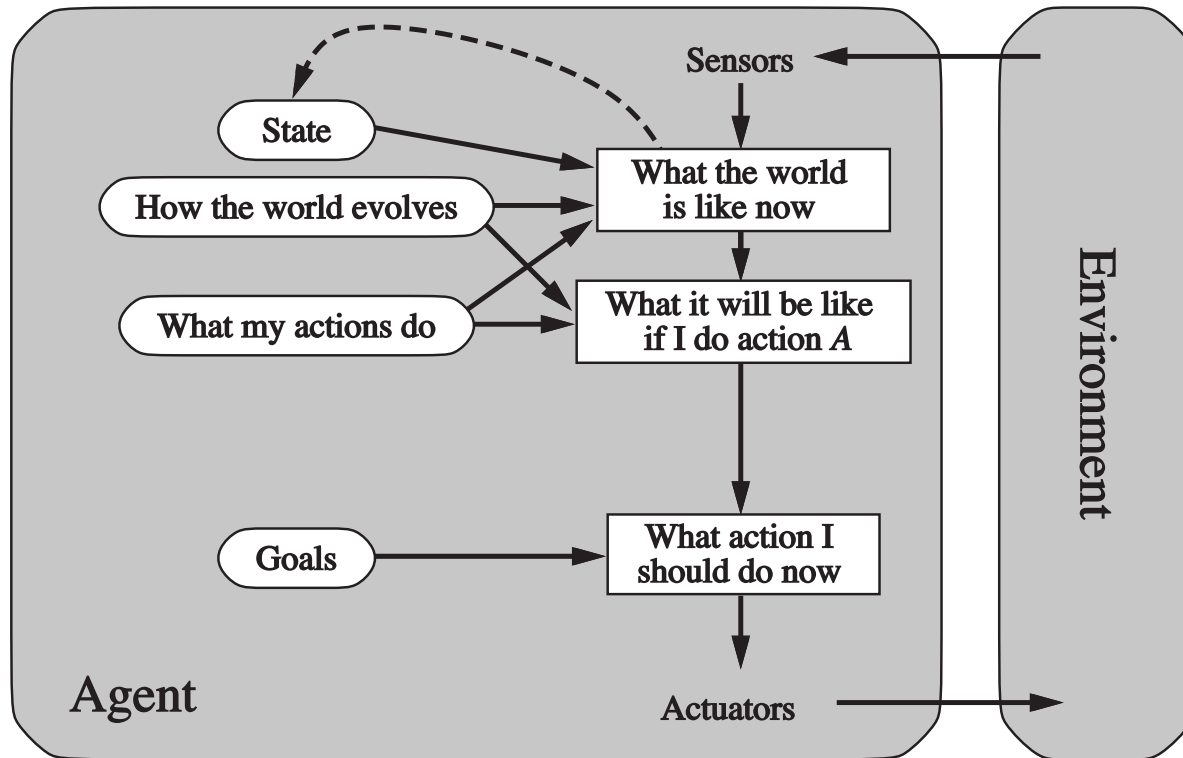
- **Simple reflex agents** select actions on the basis of the current percept, ignoring the rest of the percept history.
- They implement **condition-action rules** that match the current percept to an action.
- Rules provide a way to **compress** the function table.
  - Example (autonomous car): If a car in front of you slow down, you should break. The color and model of the car, the music on the radio or the weather are all irrelevant.
- Simple reflex agents are simple but they turn out to have **limited intelligence**.
- They can only work in a **Markovian** environment, that is if the correct decision can be made on the basis of only the current percept. In other words, if the environment is fully observable.

# Model-based reflex agents



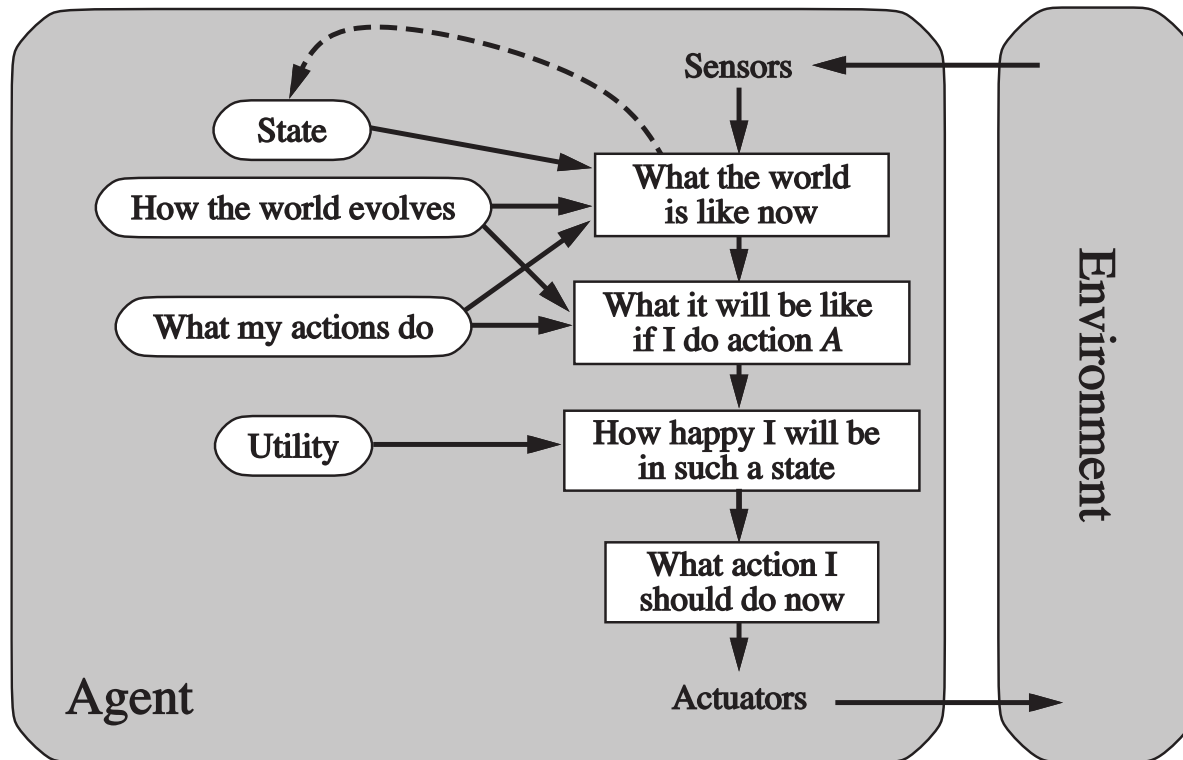
- **Model-based agents** handle partial observability of the environment by keeping track of the part of the world they cannot see now.
- The internal state of model-based agents is updated on the basis of a **model** which determines:
  - how the environment evolves independently of the agent;
  - how the agent actions affect the world.

# Goal-based agents



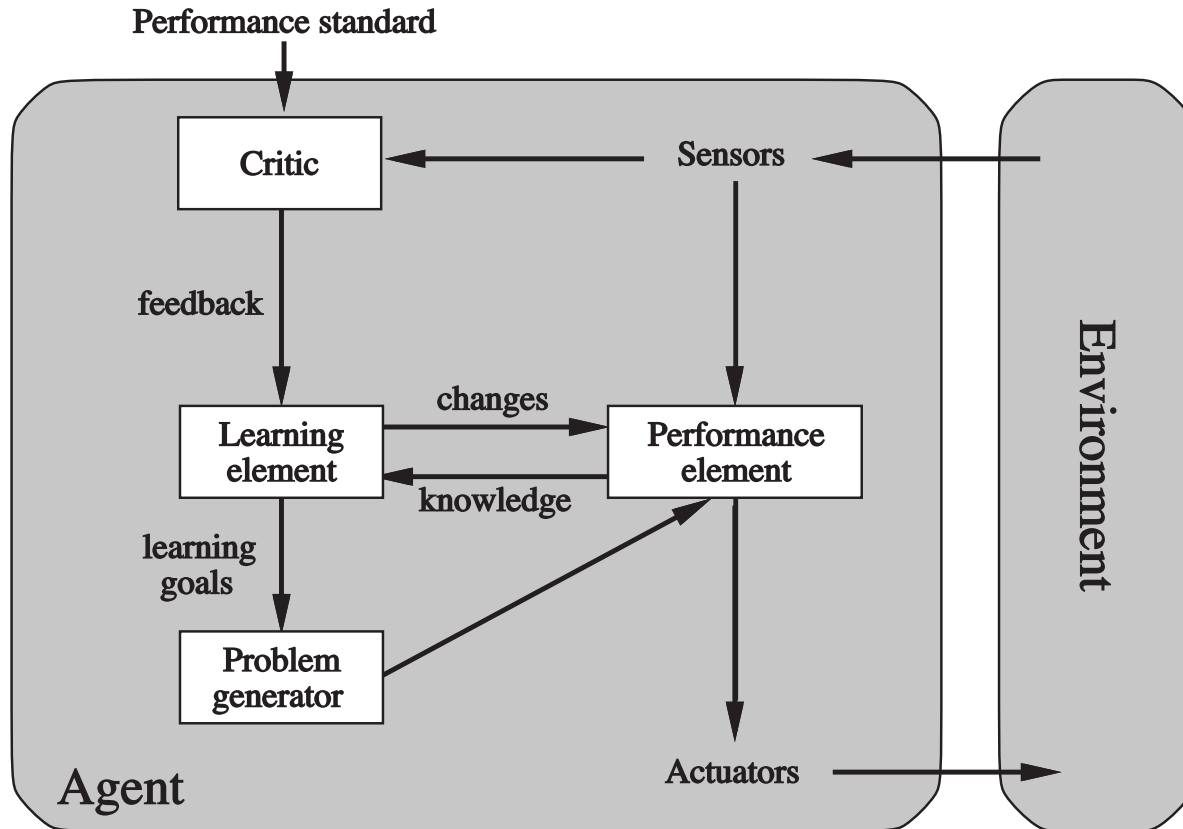
- Principle:
  1. generate possible sequences of actions
  2. predict the resulting states
  3. assess **goals** in each.
- A **goal-based agent** chooses an action that will achieve the goal.
  - More general than rules. Goals are rarely explicit in condition-action rules.
  - Finding action sequences that achieve goals is difficult. **Search** and **planning** are two strategies.
- Example (autonomous car): Has the car arrived to destination?

# Utility-based agents



- **Goals** are often not enough to generate high-quality behavior.
  - Example (autonomous car): There are many ways to arrive to destination, but some are quicker or more reliable.
  - Goals only provide binary assessment of performance.
- A **utility function** scores any given sequence of environment states.
  - The utility function is an internalization of the performance measure.
- A rational utility-based agent chooses an action that **maximizes the expected utility of its outcomes**.

# Learning agents





- **Learning agents** are capable of **self-improvement**. They can become more competent than their initial knowledge alone might allow.
- They can make changes to any of the knowledge components by:
  - learning how the **world** evolves;
  - learning what are the **consequences** of actions;
  - learning the utility of actions through **rewards**.

## A learning autonomous car

- **Performance element:**
  - The current system for selecting actions and driving.
- The **critic** observes the world and passes information to the **learning element**.
  - E.g., the car makes a quick left turn across three lanes of traffic. The critic observes shocking language from the other drivers and informs bad action.
  - The learning element tries to modify the performance element to avoid reproducing this situation in the future.
- The **problem generator** identifies certain areas of behavior in need of improvement and suggest experiments.
  - E.g., trying out the brakes on different surfaces in different weather conditions.

# Summary

- An **agent** is an entity that perceives and acts in an environment.
- The **performance measure** evaluates the agent's behavior. **Rational agents** act so as to maximize the expected value of the performance measure.
- **Task environments** includes performance measure, environment, actuators and sensors. They can vary along several significant dimensions.
- The **agent program** effectively implements the agent function. Their designs are dictated by the task environment.
- **Simple reflex agents** respond directly to percepts, whereas **model-based reflex agents** maintain internal state to track the world. **Goal-based agents** act to achieve goals while **utility-based agents** try to maximize their expected performance.
- All agents can improve their performance through **learning**.



# References

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