CS2109S: Introduction to AI and Machine Learning

Lecture 1: Intro to CS2109S and AI

18 August 2023

Intro to CS2109S

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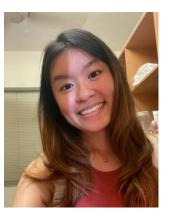
Linus Lee



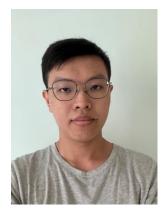
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Motivations behind CS2109S

AI/Machine learning is a big deal





Exposure to AI/ML for CS majors is deemed important by the school



Pre-survey Results

Some worries about:

- None (most popular)
- Math
 - Pre-requisites (more on this) + practice problems
- Heavy Workload
 - Follows CS2040
 - Added more scaffolding (last semester)
 - Common mistakes, e.g., using Jupyter notebook for debugging algorithms
- Difficulty
 - Not meant to be hard. Still adjusting the difficulty...
- Final Exam Format
 - We are trying to improve this

Course Pre-requisites

- CS1101S, CS1010S or equivalent
- CS1231 or equivalent
 - Trees, graphs, counting & combinatorics
- MA1521
 - Differentiation, chain rule
- CS2040S or equivalent
 - Tree and graph search
- Linear algebra
 - Vector, matrix, and their operations
- Python

- Problem Set 0

Overview



Final Assessment (2 days contest)

Week 13

Week 1 Week 5 Week 7

"Classical" Al Machine Learning **Search Algorithms ML Models & Techniques** Miscellaneous Unsupervised Uninformed search: BFS, DFS Decision Trees Learning Informed search: A* Linear/Logistic Regression • Al & Ethics Adversarial search: Minimax Support Vector Machines Neural Networks Applied CS2040S, CS1231 Python

Applied Linear Algebra, Calculus, Statistics & Probabilities
Numpy, Scikit-learn, PyTorch

"hands-on" problem sets

Textbook

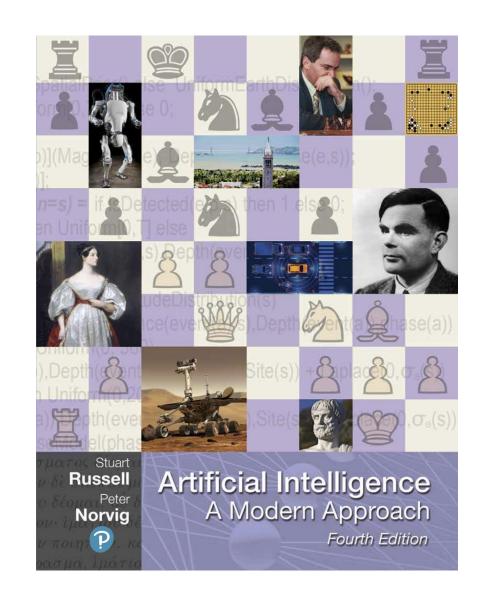
Russell and Norvig (2021)

Artificial Intelligence:

A Modern Approach (4th Ed)

a.k.a. AIMA

There is no need to buy this book.



Learning Management System



Tutorial Allocation is done through **Tutorial Survey**, <u>not</u> **Edurec**

Due Saturday, 19 Aug 23:59!



Only for **Gradebook** to record marks for checking and **webcast**

Plagiarism Policy

Dos

- Discussions without sharing/consulting/taking away any code
- Use ChatGPT with proof (e.g., give ShareGPT links)

Don'ts

- Use codes from those who has done or currently doing the course
- Use codes from the internet without proper citations
- Publish codes to any publicly accessible sites (e.g., GitHub, Google Drive) or send your codes to anyone

Plagiarism checker will be performed against all previous batches!

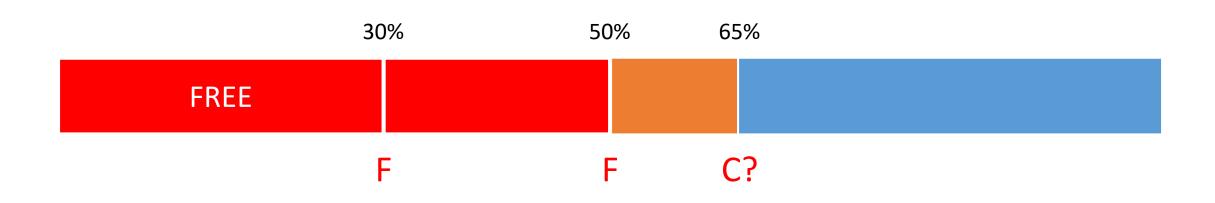
Case Studies of Plagiarisms

- Created a base source code and derived solutions from the same base
 - Caught and cases submitted to BOD
- Fully/partially copy-pasted friend's solutions (current/past student) and tried to be smart by doing some nontrivial modifications
 - Caught and cases submitted to BOD
- Fully/partially copy-pasted friend's solutions (current/past student) and claimed that solutions were from ChatGPT or ported from publicly available source codes
 - Caught and cases submitted to BOD

There were many more cases, but basically, we dealt with them appropriately

Grading

Name	Percentage
Coursemology (CA)	30% (expected: full marks)
Midterm	35%
Final Assessment	35%



Gamified CA (30%)

Survey
+50 EXP early bonus!
+100 EXP
+100 EXP

Training Tutorial

Forum EXP
+100 EXP

Training Tutorial

Forum EXP
+100 EXP

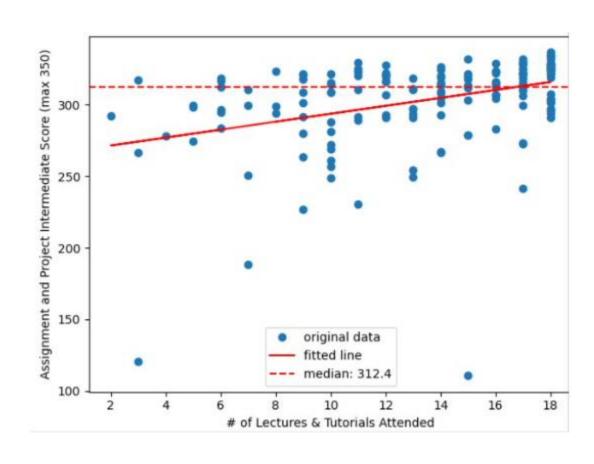
Training Tutorial

Forum EXP
+100 EXP

Background

- Enough EXP? Level Up!
- Your <u>final level</u> is your CA grade

Attendance vs Performance



Assessments

Midterm (35%)

- Closed book, open-sheet (as good as open book)
- Focus on application, not memorization

• Final (35%)

- ~2 days take-home assessment
- Focus on practical machine learning

Late Policy

- Up to 24 hours, -20%
- Up to 3 days, -30%
- Beyond 3 days, -50%

If you need extension (for valid reasons), please ask early

Intro to Artificial Intelligence

Outline

- What is AI?
- A Brief History of Al
- Intelligent Agents
- The Structure of Agents

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- What is AI?
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What is Al?









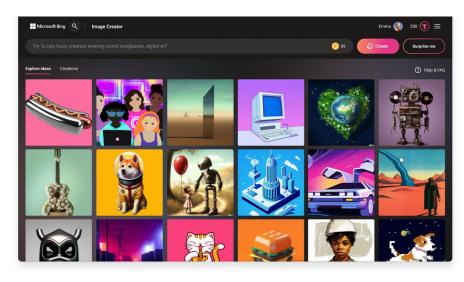
Credit: IEEE Spectrum

Credit: Guardian

Credit: NYTimes

What is Al?





Credit: Tesla Credit: Eden Al

Credit: Microsoft



What is AI? Let's ask AI

What is Artificial Intelligence in one sentence?

Artificial Intelligence is the creation of computer systems that can perform tasks and make decisions similar to human intelligence.

What is Al?

Write it as a poem*

Week 1

Week 2-4

This poem succinctly summarizes this course.

Week 5-11

Week 12

Week 13

In realms of circuits and digital might,
Artificial Intelligence takes its flight.
AlphaGo, the master of strategic plays,
DeepBlue conquers chess in astonishing ways.

Watson, the genius with knowledge immense, ChatGPT converses with eloquence. Self-driving cars, navigating with grace, Image generation, artistry embraced.

Search algorithms, exploring vast terrain,
Decision trees, guiding paths to gain.
Linear models, simple and clear,
Neural networks, complexity they steer.

Unsupervised learning, a quest for patterns,
Unveiling insights, where hidden knowledge hammers.
Ethics guide us in this AI realm,
Respecting privacy, and doing no harm.

Artificial Intelligence, a powerful force,
With potential vast, both boon and discourse.
May we harness its strength with a watchful eye,
To shape a better future as days roll by.



^{*)} This is not the actual (full) prompt

Outline

- What is AI?
- A Brief History of Al
- Intelligent Agents
- The Structure of Agents

A Brief History of Al

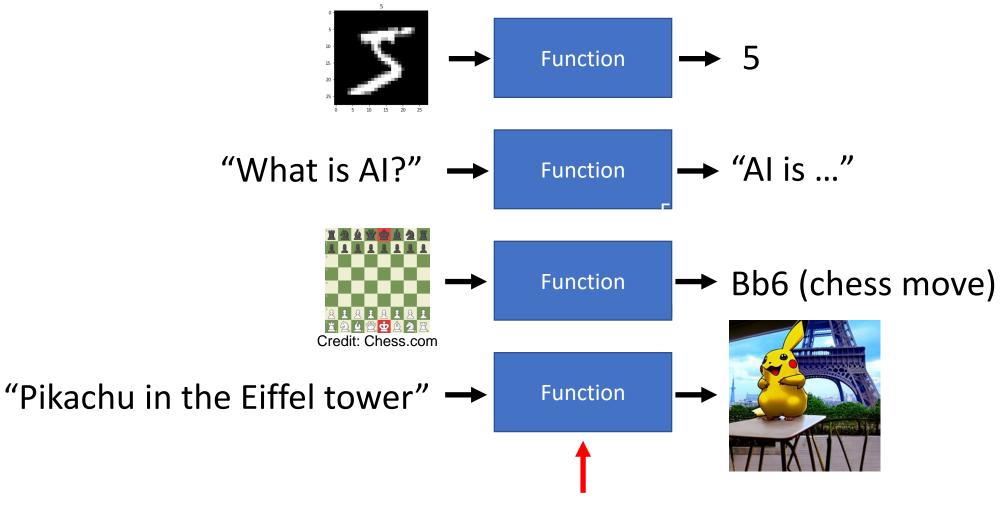
- Before 1900
- Early 1900
 - Atanasoff-Berry Computer: solve linear equations
 - Neural networks
- 1950s-1960s
 - Alan Turing: Turing machines, Turing test
 - Checkers AI, Lisp programming language, ELIZA
- 1970s
 - 1st Al Winter
- 1980s
 - Expert Systems, Fifth Generation Computers
- 1990s
 - Deep Blue
- 2000-2010
 - 2nd Al Winter, Big data
- 2011-Beyond
 - Deep neural networks, Watson, AlphaGo, ChatGPT

Outline

- What is AI?
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Intelligent Agents: 1st Attempt

- Handcraft → "Classical" AI
- Learn \rightarrow Machine Learning



How do we write this Function?

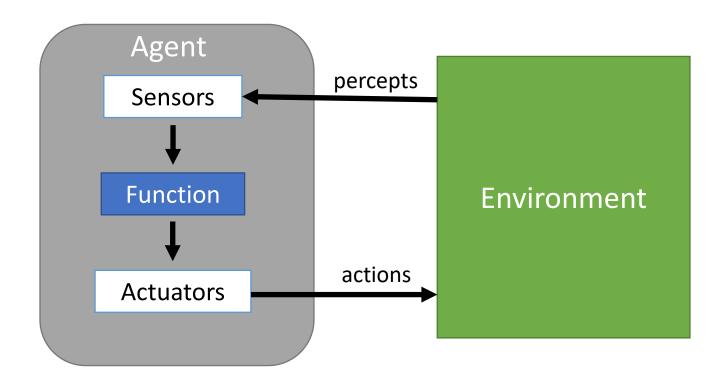
Intelligent Agents: 1st Attempt

Real-world problems are usually not that simple



It usually requires a feedback loop between the AI and the world

Intelligent Agents

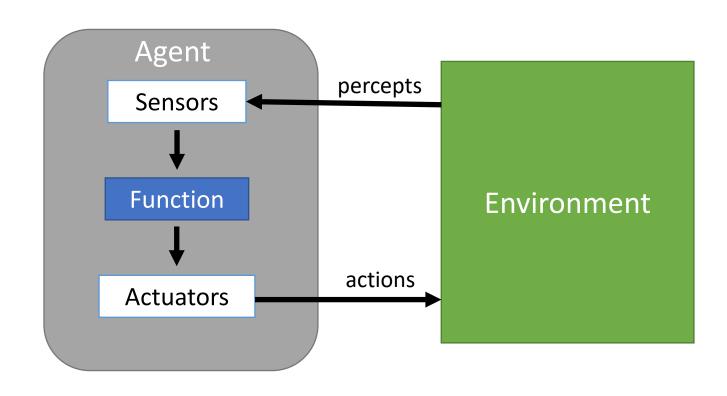


The agent function maps from percept histories to actions:

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

The agent program runs on the physical architecture to produce function f

Intelligent Agents How would agent know to do the right thing?

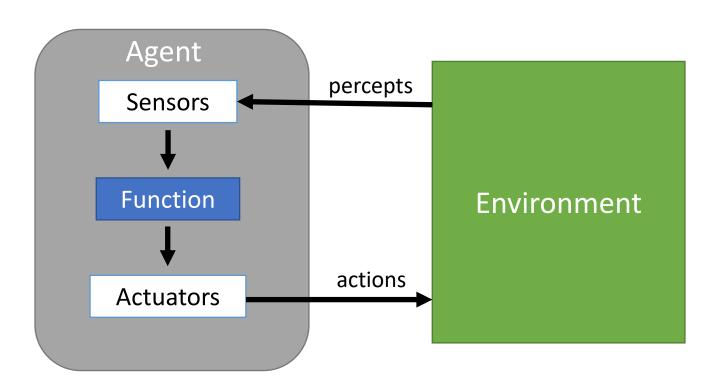


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$$[f: \mathcal{P}^* \rightarrow \mathcal{A}]$$

The agent program runs on the physical architecture to produce function f

Intelligent Agents How would agent know to do the right thing?



Performance Measure

Things to consider:

- Best for whom?
- What are we optimizing?
- What information is available?
- Any unintended effects?
- What are the costs?

A rational agent will choose actions that maximize performance measure.

PEAS

Performance Measure, Environment, Actuators, Sensors

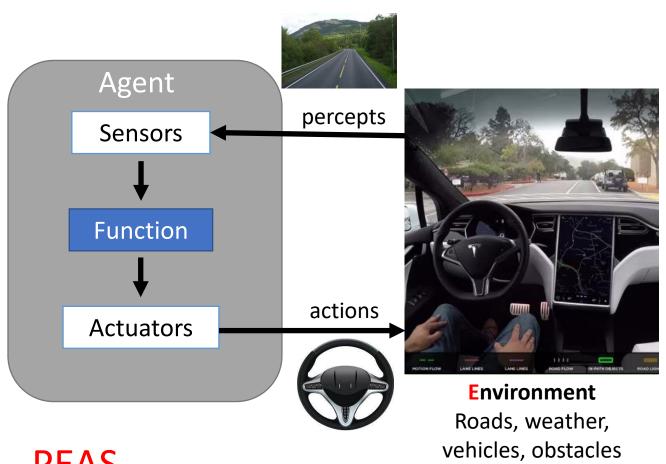
Intelligent Agents: Autonomous Driving

Sensors

- Camera
- LIDAR
- Speedometer

Actuators

- Steering wheel
- Accelerator
- Brake



Performance Measure

- Safety
- Speed
- Legal
- Comfort

PEAS

Performance Measure, Environment, Actuators, Sensors

Intelligent Agents: AI Assistant/Chatbot

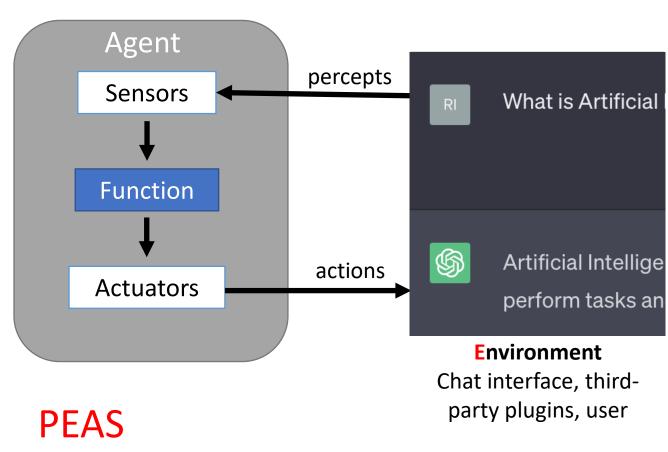
Sensors

- Text input
- Chat history
- Context
- ...

Actuators

- Text output
- Image output
- API

• ...



Performance Measure

- Correctness
- Conciseness
- Legal
- Safety
- ...

Performance Measure, Environment, Actuators, Sensors

Properties of Task Environment

Full



Strategic



Deterministic

Partial



Stochastic/Strategic

Fully observable (vs. partially observable)

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

Deterministic (vs. stochastic)

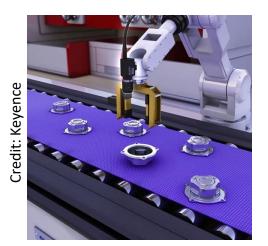
The next state of the environment is completely determined by the current state and the action executed by the agent. (If the environment is deterministic except for the actions of other agents, then the environment is **strategic**)

Properties of Task Environment

Sequential



Static/Semi-Dynamic



Episodic

Sequential



Dynamic

Episodic (vs. sequential)

The agent's experience is divided into atomic "episodes" (each episode consists of the agent perceiving and then performing a single action), and the choice of action in each episode depends only on the episode itself.

Static (vs. dynamic)

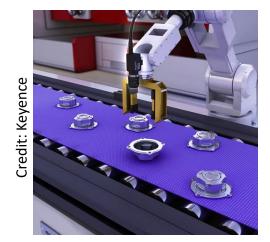
The environment is unchanged while an agent is deliberating. (The environment is semi-dynamic if the environment itself does not change with the passage of time, but the agent's performance score does)

Properties of Task Environment

Discrete



Multi-agent



Single-agent

Continuous



Multi-agent

Discrete (vs. continuous)

A limited number of distinct, clearly defined percepts and actions.

Single agent (vs. multi-agent)

An agent operating by itself in an environment.

Outline

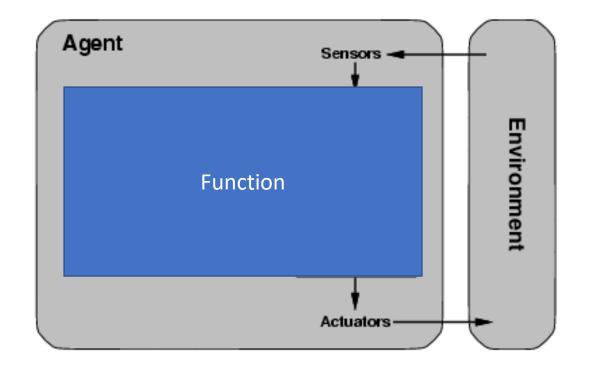
- What is AI?
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The Structure of Agents

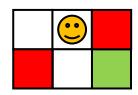
An agent is <u>completely specified</u> by the **agent function** mapping percept sequences to actions.

Common agent structures

- Simple reflex agents
- Model-based reflex agents
- Goal-based agents
- Utility-based agents
- Learning agents



Simple Reflex Agent



Condition-action rule

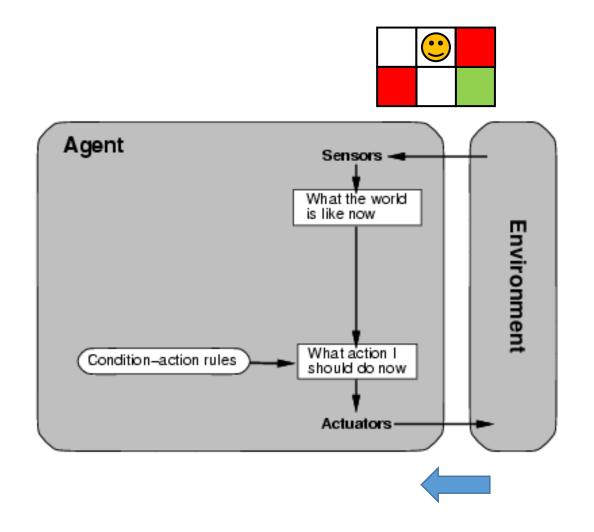
If left empty: left

If right empty: right

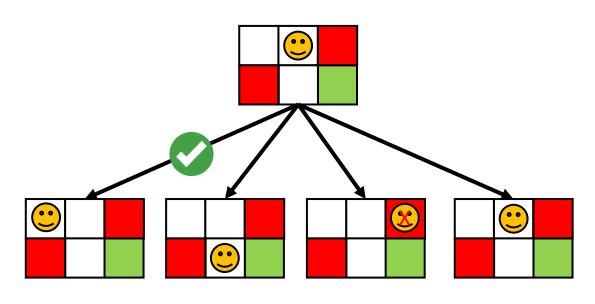
If up empty: up

If down empty: down



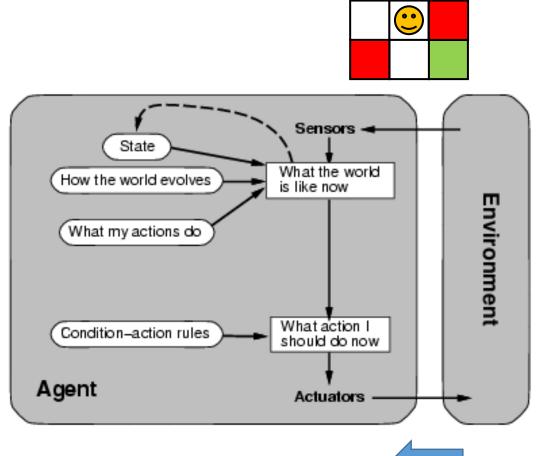


Model-based Agent

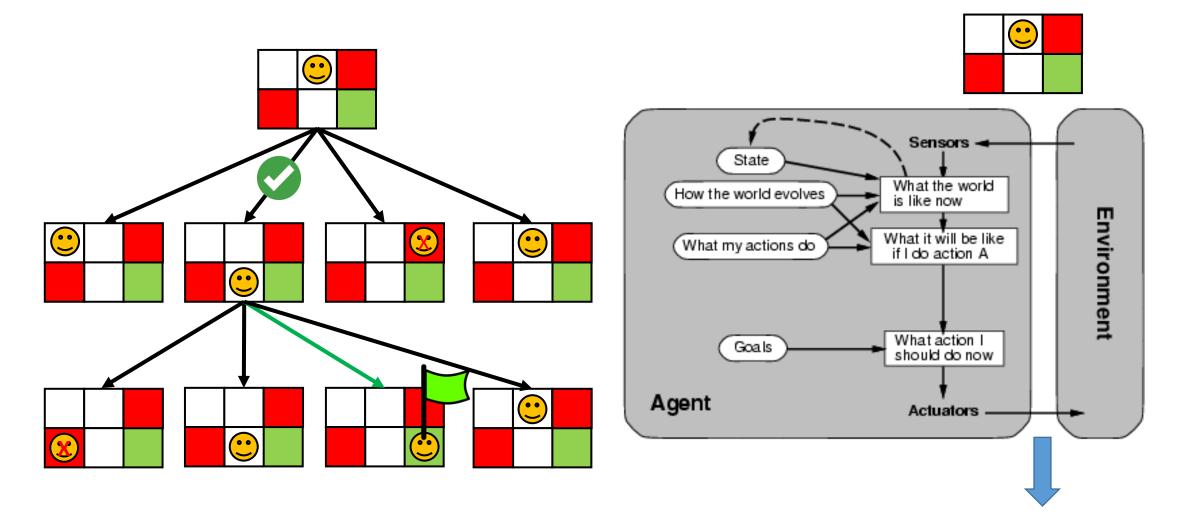


Condition-action rule

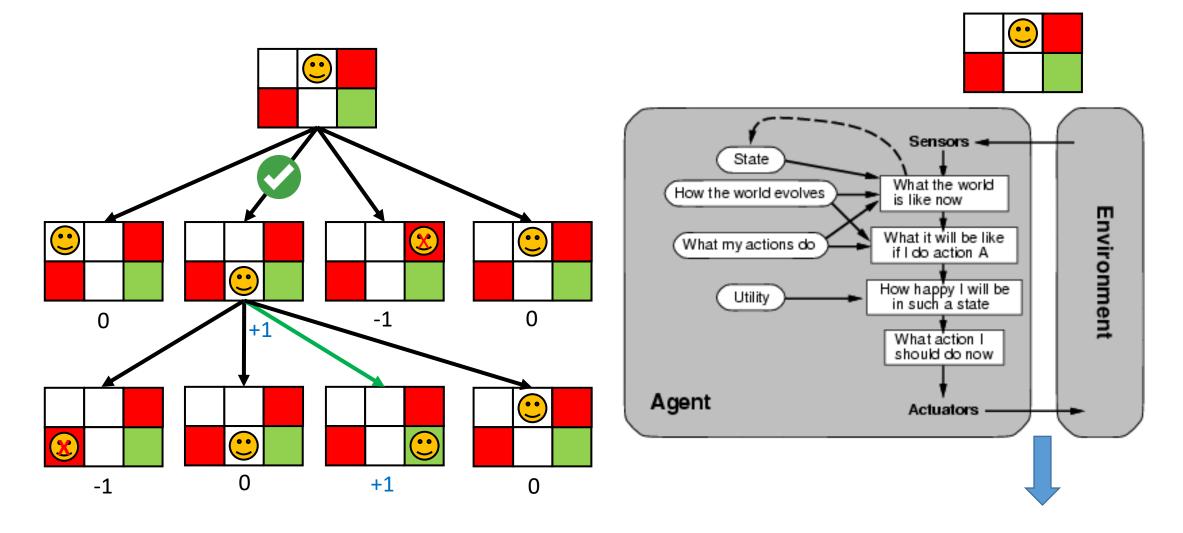
If not die, pick the action



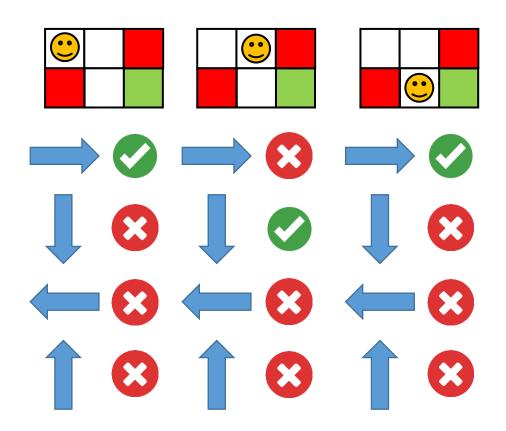
Goal-based Agent

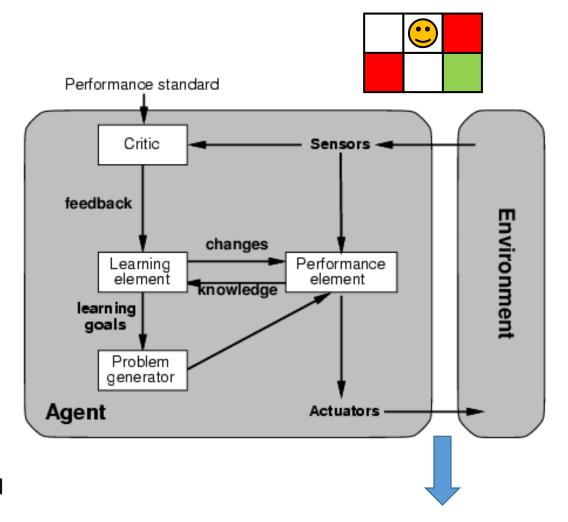


Utility-based Agent



Learning Agent





Can be reflex, model-based, goal-based, and utility-based

Exploitation vs Exploration

An agent operating in the real world must often choose between:

 Maximizing its expected utility according to its current knowledge about the world (Exploit)

Trying to learn more about the world (Explore)



VS



Credit: Amazon

Summary

- Al: computers trying to behave like humans
- **PEAS** Framework:
 - Performance measure: define "goodness" of a solution
 - Environment: define what the agent can and cannot do
 - Actuators: outputs
 - **S**ensors: inputs
- Agent function is <u>sufficient</u> to define an Al agent
- Common agent structures:
 - Reflex, model-based, goal-based, utility-based, learning
- Exploration vs exploitation

Coming Up Next Week

- Formulating search problems
- Uninformed search algorithms
 - Breadth-first search
 - Depth-first search
 - Uniform-cost search
- How to handle repeated states?
 - Memoization / graph search
- How to handle infinite depth in search?
 - Depth-limited search
 - Iterative deepening search
- Forward, Backward, Bidirectional search

To Do

- Lecture Training 1
 - +100 Free EXP
 - +50 Early bird bonus
- Problem Set 0
 - Due 26 August
- Practice Problems: Python and Numpy (optional)