Introduction to Computer Science Lecture 10: Artificial Intelligence

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What is AI?

Cognitive Science

LogicistIs that human?

Think like humans	Think rationally
Act like humans	Act rationally

Turing (1950) Test

Eliza: chayden.net/eliza/Eliza.html www.jabberwacky.com

Natural language processing Knowledge representation Automated reasoning Machine learning Rational Agent

Strong Al vs. Weak Al

- Weak Al
 - Machines can be programmed to exhibit intelligent behavior.
- Strong Al
 - Machines can be programmed to possess intelligence and consciousness.



John Searle's Chinese room argument.

Levels of Intelligent Behaviors

- Reflex: actions are predetermined responses to the input data
- More intelligent behavior requires knowledge of the environment and involves such activities as:
 - Goal seeking
 - Learning

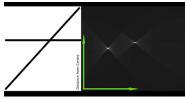
Research Approaches in Al

- Performance oriented
 - Engineering track
 - To maximize the performance of the agents.
- Simulation oriented
 - Theoretical track
 - To understand how the agents produce responses.

Understanding Images

- Template matching
- Image processing
 - edge enhancement
 - region finding
 - smoothing
- Image analysis
 - Hough transformation (line, circles)





Natural Language Processing

- Syntactic analysis
 - Mary gave John a birthday card.
 - John got a birthday card.

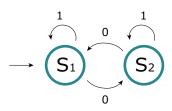
- Subject: Mary Subject: John
- Semantic analysis, contextual analysis
 - John drove me home.
 - John drove me crazy.
 - The pigpen was built by the barn.
 - The pigpen was built by the farmer.
 - Do you know what time it is?

Natural Language Processing (contd.)

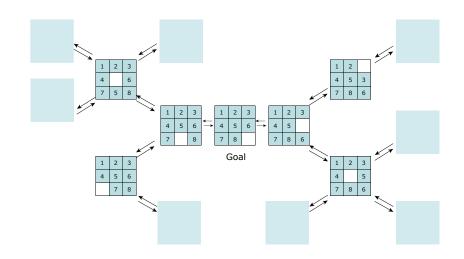
- Information retrieval / extraction
 - I've got a solution to your problem.
 - Shoot.
 - Right.
 - How was your date last night?
 - He/She has a good personality.
 - You can count on me.
 - Ya, right. That's comforting.

Reasoning

- Production systems
 - Collection of states including initial state & goal state(s)
 - Collection of productions: rules or moves
 - Each production may have preconditions
 - Control system: decides which production to apply next
- Recall prolog
- Similar to finite state automata



Search a Production System

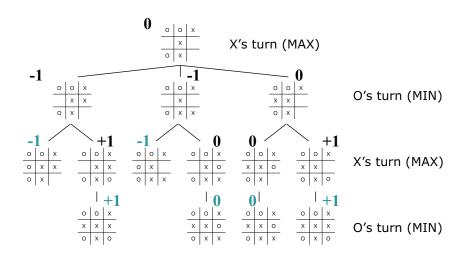


Computer Game Playing

- Let's meet an old friend
 - Tic-tac-toe



Game Tree & Minimax Search



Heuristic

- For most games, a complete search is practically impossible.
 - Chess $\sim 10^{47}$; Chinese chess $\sim 10^{48}$; Go $\sim 10^{171}$
- A quantitative estimate of the distance to a goal is needed.
- Requirements for good heuristics
 - Much easier to compute than a complete solution
 - Reasonable estimate of proximity to a goal

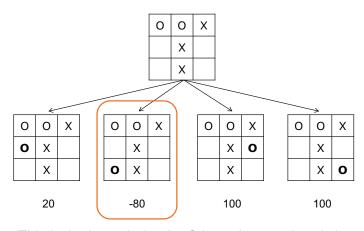
Let's Define a Heuristic

XX_{-}	100
X	10
	0
OX*	0
0	-10
00_	-100

0	0	×
	×	
	×	

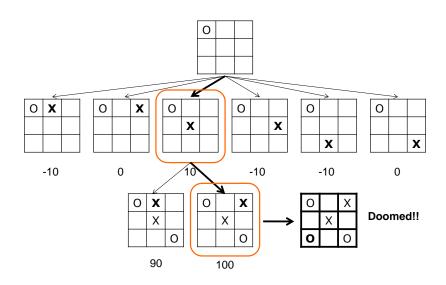
$$\longrightarrow \begin{array}{c} 0+10+10-10+0+10+0+100 = 120 \\ \text{The board favors X} \end{array}$$

Does It Work?



This is the best choice for O based on our heuristic.

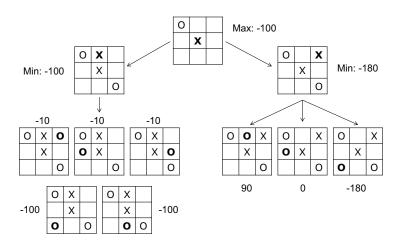
How About This?



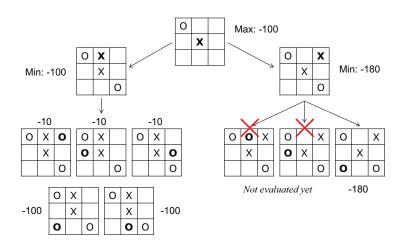
What's Wrong?

- Heuristics are not perfect
 - Otherwise, we'd call them solutions
- Heuristics are usually more accurate toward the end of the game.
- Need some search procedure for more accurate estimation.

Heuristic + Minimax Search



Alpha-Beta Pruning

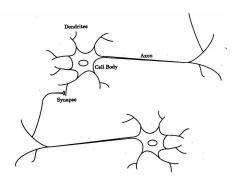


Learning

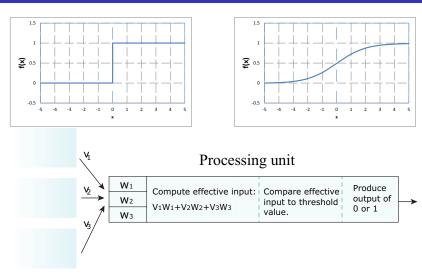
- Supervised vs. unsupervised
- Supervised
 - Learning by provided examples
 - Imitation
 - Parameter tuning
- Unsupervised
 - Learning by experiences
 - Reinforcement
 - Evolutionary (semi-supervised)

Artificial Neural Networks

- Human brain
 - 10¹¹ neurons
 - 10¹⁴ synapses



Perceptron



1 iff greater than or equal to the threshold

Some Building Blocks

AND

$$\begin{array}{cccc}
x & \longrightarrow & \boxed{1} \\
y & \longrightarrow & \boxed{1}
\end{array}$$
 1.5

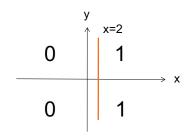
OR

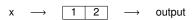
$$\begin{array}{cccc} x & \longrightarrow & \boxed{1} \\ y & \longrightarrow & \boxed{1} \end{array}$$
 0.5

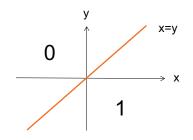
SIGN

$$x \longrightarrow \boxed{1} \boxed{0} \boxed{-}$$

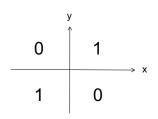
Some Examples

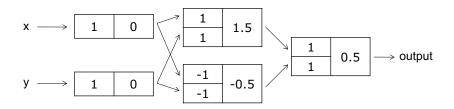






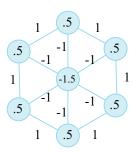
The XOR Problem





Associative Memory

Content addressable



How Does It Work

a.



Start: All but the rightmost units are excited

c.



Step 2: The top and bottom units become excited

b.



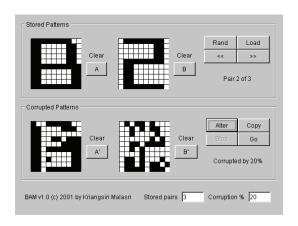
Step 1: Only the leftmost units remain excited

d.



Final: All the units on the perimeter are excited

Example



BAM applet: http://www.cbu.edu/~pong/ai/bam/bamapplet.html

Darwin's Theory of Evolution

- Evolution
 - The change in populations of organisms over generations.
- Darwin's idea: Natural selection
 - Struggle to survive
 - Survival of the fittest
 - Genetic variation: inherited traits



Black-Box Optimization



- Finding the x that yields the highest y with an unknown f
- Evolving the giraffe that is the fittest in an unknown environment.
- Instead of finding a solution, let's evolve a solution.

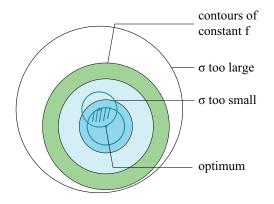
(1+1) Evolutionary Strategy

- Simplest evolutionary strategy
- One parent: n-dimension real vector, $P = (p_1, ..., p_n)$
- Generate one child by mutation: $C = (c_1, ..., c_n)$
 - $-c_i=p_i+N(0,\sigma^2)$
- Replace P by C if C is better.
- Modify σ according to the replacement rate r.
 - One fifth rule

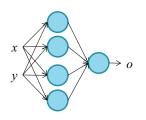
1/5 Rule Intuition

- $\sigma \leftarrow \sigma/C^{1/n}$, if $r > \Theta$
- $\sigma \leftarrow \sigma \cdot C^{1/n}$, if $r < \Theta$
- If replacement rate high, not exploring enough → increase step size.
- If replacement rate low, too daring → reduce step size.
- ullet $\Theta=1/5$ (Guessed by Rechenberg) and C=0.817 (Progress analysis by Schwefel)

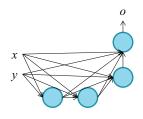
Visualization of 1/5 Rule



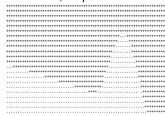
Training NN with (1+1)ES



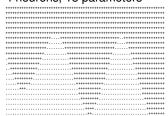




5 neurons, 17 parameters



4 neurons, 18 parameters



Total Differences: 172(123.084600)/1540