Natural Computation and Behavioral Robotics

Genetic versus Adaptive Intelligence

Overview

- What is Genetic Intelligence?
- What is Adaptive Intelligence?
- How does each one produces "intelligence"?
- How does "memory" affects each one?
- How does "knowledge" is encoded in them?
- When should we prefer to use each one?

Genetic Intelligence

- A property is hard-coded in a DNA sequence that constitutes a gene or "genotype"
- Each gene corresponds to a different functionality or a "phenotype" (result when "decoding" each gene)
- Genes are <u>mutated</u> and <u>combined</u> during a reproduction process (⇒ combined "functionalities")
- Best (combined) functionalities <u>survive</u>, i.e. their corresponding genes get to survive too
- Multiple mutation-selection cycles produce optimal populations (=> best-fitted solutions to a problem)

Genetic Programming

Basic idea:

- Encode a problem's parameters into artificial "genes" of encoded information (strings of bits or characters)
- Implement software functions for "mutation" and "selection" steps of evolution cycle
- 3. Set a large pool of "genes" in some initial (random) values and let the population "evolve" towards an optimal solution after a large number of cycles

Implementing Genetic Algorithms

- <u>Initialize</u>: produce a large number of genes with random values as the starting population
- <u>Crossover</u>: take two genes and use one part of the first and the remaining part from the second to produce a new "offspring"
- <u>Mutation</u>: change a specific point in a single gene, from one value to another
- <u>Fitness</u>: compute how well a "gene" works for the current environment
- <u>Update</u>: eliminate "bad" genes and keep the rest for the next evolution cycle (⇒ "selection" process in Nature)

Adaptive Intelligence

Basic idea:

- Use a very "dumb" and very low-level functionality as the starting base
- 2. Use a small or large "<u>memory</u>" pool to store previous action-result associations as "past experience" or "knowledge"
- 3. Exploit gained "knowledge" to constantly "adapt" and improve future decisions

Adaptive Systems

- They start off with no experience or "<u>hard-</u> <u>coded</u>" knowledge about their environment
- First few decisions are usually very "bad", but they "learn" to behave better over time
- Actions leading to good results are preferred over actions leading to bad results
- Over time, they <u>develop</u> optimal behavior for the <u>current</u> environment (⇒ "survive")

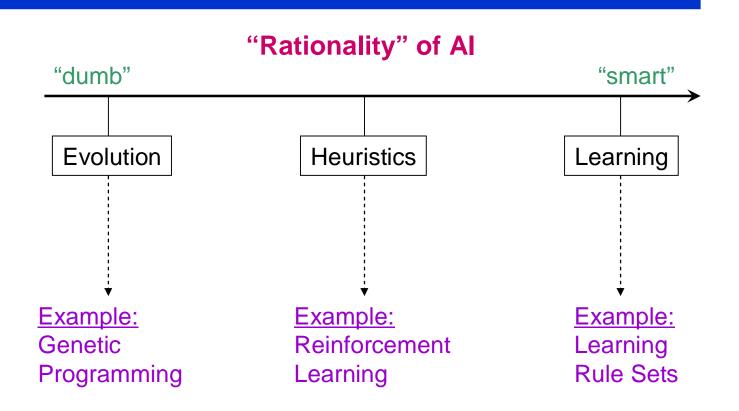
Implementing Adaptive Behavior

- Generic approach: "Reinforcement Learning"
- Implement adaptive decision-making by employing an <u>iterative update rule</u> for "gain"
- Adaptiveness includes:
 - "Reference Points": past cases of good decisions
 - "Experimentation": how often to ignore experience
 - "Forgetfulness": how quickly to forget experience
- Although simple, they often result in erratic or even chaotic behavior (see: logistic equation, fractals)

Genetic versus Adaptive Agents

	Genetic Al	Adaptive Al
Start-off:	Large population	Small population
Optimize:	Evolution of "good" genes	Learn from their mistakes
Advantages:	Effective for static environments	Adapt to dynamic environments
Pitfalls:	Genes do not become "smarter"	Must survive until they get "smart"

Degrees of Rationality



Food for thought

- When should we use Genetic-based Al?
- When should we use Adaptive-based AI?
- Do animals use Genetic or Adaptive AI?
- How does the "memory" factor is <u>implemented</u> in each one of them ?
- How does this "memory" affects and/or propagates the optimality in each case?

P.C. – Readings

- Tom Mitchell, "Machine Learning", McGrawHill, 1997. [see: ch.9, ch.13]
- John L. Casti, "Reality Rules II: Picturing the World in Mathematics – The Frontier", John Wiley & Sons, 1997. [see: ch.7]
- Sergui Hart, "Adaptive Heuristics", Journal of Economic Literature, Dec.2004
- Mikhael Shor, "Learning to Respond: The Use of Heuristics in Dynamic Games", Vanderbilt Univ., Jun.2004