

Automated Life Ecosystem Simulator

Artificial Intelligence and Adaptive Systems

John P. Daigle

Department of Computer Science
Georgia State University

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Outline

- 1 Artificial Intellingence Paradigms
 - Classical Approach
 - Analogical Engine
 - Adaptive System
- 2 Artifical Life
 - Emergent Behavior and Complexity
- 3 ALES
 - History and Goals
 - Current System
- 4 Conclusion

expert system

- Implementations
 - An engine connected to a database
 - A domain specific knowledgebase
 - Can use probability, Neural Networks, etc.
 - Works well in medicine, online help systems
- Limitations
 - domain specificity: the less flexible the system, the more “expert”
 - knowledge discovery: expert systems typically don’t make novel connections
 - flexibility: expert systems typically are not adaptive

general concept

Difference Engine A mechanical calculator, no memory.

Analytical Engine A programmable calculator with a memory

Turing Machine Formal model of a Analytical Engine

Analogical Engine A machine capable of isomorphisms

Warning!

This is not a standard term!

Isomorphism A mapping from one domain to another

Example

What is the Atlanta of China?

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understanding analogies

- Why are analogies Interesting?
 - Knowledge acquisition is difficult
 - Analogies provide context between multiple domains
- Analogies provide flexibility
 - An expert system provides an answer based on prior knowledge of the current domain
 - An analogical system could guess, based on prior knowledge of different domains

current research

Metacat

Metacat is a microdomain AI that attempts to build analogies. The metacat microdomain is the roman alphabet, and it uses concepts such as repetition, succession, length, and so forth to solve puzzles of the form

$$aabb \rightarrow bbcc; eeff \rightarrow? \quad (1)$$

LetterSpirit

Letter spirit attempts to create novel grid fonts.

- Very fringe area
- gave rise to very popular area: emergent behavior

a short history

- John Holland
 - First paper on Adaptation 1962
 - Holland Classifiers 1976
 - Complex Adaptive Systems remain interesting today
- Trends
 - Originally, it was thought that a central controller was needed
 - This “model based approach” was popular until recently
 - Replaced by an agent based approach, no central control
 - To some degree, this is what we are studying

philosophic

Theorem

Clearly, if adaptive problem solving is a field of AI, ALife is a field of AI. Life practically is adaptive problem solving, at the level of agent, ecosystem, species, and even DNA molecule.

emergent behavior

Emergent Behavior A complex or unpredictable result from a simple set of rules

- Flocking or Schooling Organisms
- Stars, weather systems
- Economies

Discrete Event Simulation A method of simulation that attempts to model the behavior of each agent or object in a system, allows for emergent rather than averaged results.

- more accurate results for complex systems
- only approach that allows adaptive behavior

swarms in computing

SWARM

SWARM is a software program built at the Santa Fe Institute, a multidisciplinary research institutio. SFI studies complexity, complex systems, and adaptation.

- A swarm intelligent system creates solutions to problems through modeled behavior
- Can be used to solve difficult problems
- TSP in polynomial time with an ANT simulator[7]
- Ad hoc network creation and maintenance[14]

swarms in nature

- Not always homogenous
- immune system is considered intelligent swarm[8, 12, 11]
- can exist in multi-tiered relationships

Fertile Ground for Thought

My Dog is a lot more intelligent than Deep Blue.

Foster	Deep Blue
Many Tricks	One Trick
Arbitrary Commands	Delineated Commands
Flexible Goals	Rigid Goals
Self Directed	Other Directed

But isn't he just a collection of Swarms? Aren't I?

what is ALES

The Automated Life Ecosystem Simulator

The ALES was a software engineering project at Georgia State University in the Fall of 2005.

Design Team Myself, Ed Bullwinkel, Larry Eisenstein, Michael Balaun.

Design Goal A self sustaining artificial ecosystem for interactive use

Description ALES emulates single celled life forms in a watery environment. The feedback between different creatures and the environment creates an adaptive system in its totality.

state of ALES in December, 2005

When the ALES project terminated, we had a system that would run on exactly one laptop computer, after being compiled from netbeans. It contained three creatures, but these creatures only had a single behavior each. The user had to “prop up” the environment or the simulation would terminate quickly.

philosophic

Goals

- Create sourceforge project, to allow anyone access to the code
- incorporate genetic algorithms
- create feedback
- increase behaviors

At this time, the ALES creatures do not evolve. Because they were not designed too, it has proven difficult to create a set of rules by which traits would be produced.

simulation behavior

Game Board has oxygen and nutrients

Yellow Algae reduces nutrients, increases oxygen

Terramecium reduces oxygen, eats algae

Death increases nutrients

The theory is that the emergent behavior of these simple interactions will be a stable, self regulating system!

recap





- Traditional AI lacks facilities for Isomorphism
- Living systems are able to create analogies
- Traditional AI lacks adaptability
- Living systems are Adapatable

Living system simulation will enhance our understanding of intelligence and expand what is possible with computers. Understanding emergent behaviors of complex systems is the next frontier for AI research.[2]




ALES falls short

- The current ALES system is not intelligent
 - ALES is adaptable
 - ALES creatures exhibit no Swarm behaviors
 - ALES needs an inheritance model
- Future ALES systems will be more complex
 - Immune responses
 - more creature types
 - localized environment for more complex information propagation



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