

## AI Exotics

### Conventional AI Topics in 1999

knowledge representation and articulation  
commonsense and uncertain reasoning, belief structures  
inference and deduction, expert systems  
search (heuristics, strategies, competitive)  
blackboard systems  
learning and adaptation  
deliberation, planning, and acting  
situated modeling  
speech and language processing  
image understanding and synthesis  
manipulation and locomotion (robotics)  
autonomous agents and robots  
multiagent systems  
cognitive modeling  
mathematical foundations

### Probable Directions of Research over the next few years

representing the environment as messy, changing, and hostile  
huge knowledge bases, data mining  
interaction with human collaborators  
system self-understanding  
self-motivated learning and agents  
more agents and multiagent systems (the next big thing)

### Exotic Topics in AI

neural networks (1985 old, has own technical societies)  
multiagent systems (becoming mainstream)  
autonomous agents, artificial life  
data mining, particularly in dynamic web environments  
fuzzy logics (1990 old)  
constraint reasoning (1980 old)  
chaotic systems, cellular automata (1980 old)  
genetic algorithms (1990 old)  
machine evolution (1996)  
reactive machines (1990 old)  
robot perception (1980 old but little progress)  
diagrammatic systems (1990)  
web intelligence

## Neural Networks

Densely connected networks with input, middle, and output layers

Inherently parallel architecture

Good for pattern-recognition after training

Sub-symbolic (no map between representation and meaning)

Knowledge (and programming) is in the strength of connections between network units

Each middle unit "averages" inputs, eg:  $outK@t = \text{Sum}(\text{weight}J * inJ@t) + \text{bias}K$

Weighted summation is a discrete model of differential calculus

Logistic summation:  $out@t = 1 / (1 + e^{-(\text{sum}(\text{weight}J * inJ))})$

simulates perceptual threshold model

Training is accomplished by modifying weights to meet a performance objective

Programming is reflexive,

ie the system reprograms its own weights through error feedback

Open issues:

How to learn to solve a problem without training in the exact problem?

Are there other useful architectures (compared to three layer approach)?

Can training be less extensive for hard problems; does training scale up?

How can a network generalize to a class of problem when trained in a subset?

How can you know the state of the system's knowledge? (validation issue)

Experience cautions:

The training examples must sufficiently constrain the problem.

Selection of appropriate input data is a design problem.

Known symmetries must be included in the training set.

The task needs a probabilistic model.

Don't train with binary vectors (errors are fatal)

Course coding (information blurring) is good for pre-training

## Cellular Automata

"discrete dynamical systems whose behavior is completely specified in terms of a local relation, much as is the case for a large class of continuous dynamical systems defined by partial differential equations" Tommaso & Toffoli

Simple example is domino runs.

Highly parallel recursive technique

Complex behavior (or is it?) from simple rules

Primary example of self-generating infinite systems

Most simple automata are Turing equivalent

Outcome is not "predictable", since running the automata is the only computational model

Main result: Complexity occurs at phase transitions between simple and random

## Agents

Software agents are programs with "attitude", or disposition

Agents act autonomously in support of a user's requirement

Programming techniques for agents include

## Artificial Intelligence

- reactivity (selective activity depending on context)
  - autonomy (goal directed and self-initializing activity)
  - collaboration (attunement to work with other agents)
  - knowledge-based (use rules for communication protocols)
  - inferential (deduce consequences from inputs)
  - temporal (persistence of state and identity)
  - personality (believable interface with emotional simulations)
  - adaptivity (self-changing over time)
  - mobility (migrate to different environments, usually in search)
- Purposes of agent approaches
- simplify distributed computing
    - automated interoperability
    - resource management
  - improve user interface
    - simulate more human activity
    - relieve burden of direct interface
    - indirect management (vague specification)
  - system architecture

## Genetic and Evolutionary Algorithms

- Genetic algorithms are designed using "natural selection" and random permutation. Code fragments or bit-strings are modified by
- selection (those meeting a criteria are strengthened)
  - crossover (two fragments swap pieces at a cut point)
  - mutation (parts of a fragment are changed, usually randomly)
- Millions of cycles are required for algorithm growth or evolution  
Art is a good application.

## Artificial Life

- Self-replicating programs (viruses)
- True autonomy ("Sorry, Dave, I can't let you do that.")
- Graphics for Hollywood
- Animal construction kits
- Cellular automata, chaos, and fractal mathematics
- Issues:
  - physical grounding hypothesis (representation is not needed)
  - What is autonomy? What is emergent behavior?
  - What is the relationship between behavior and environment?