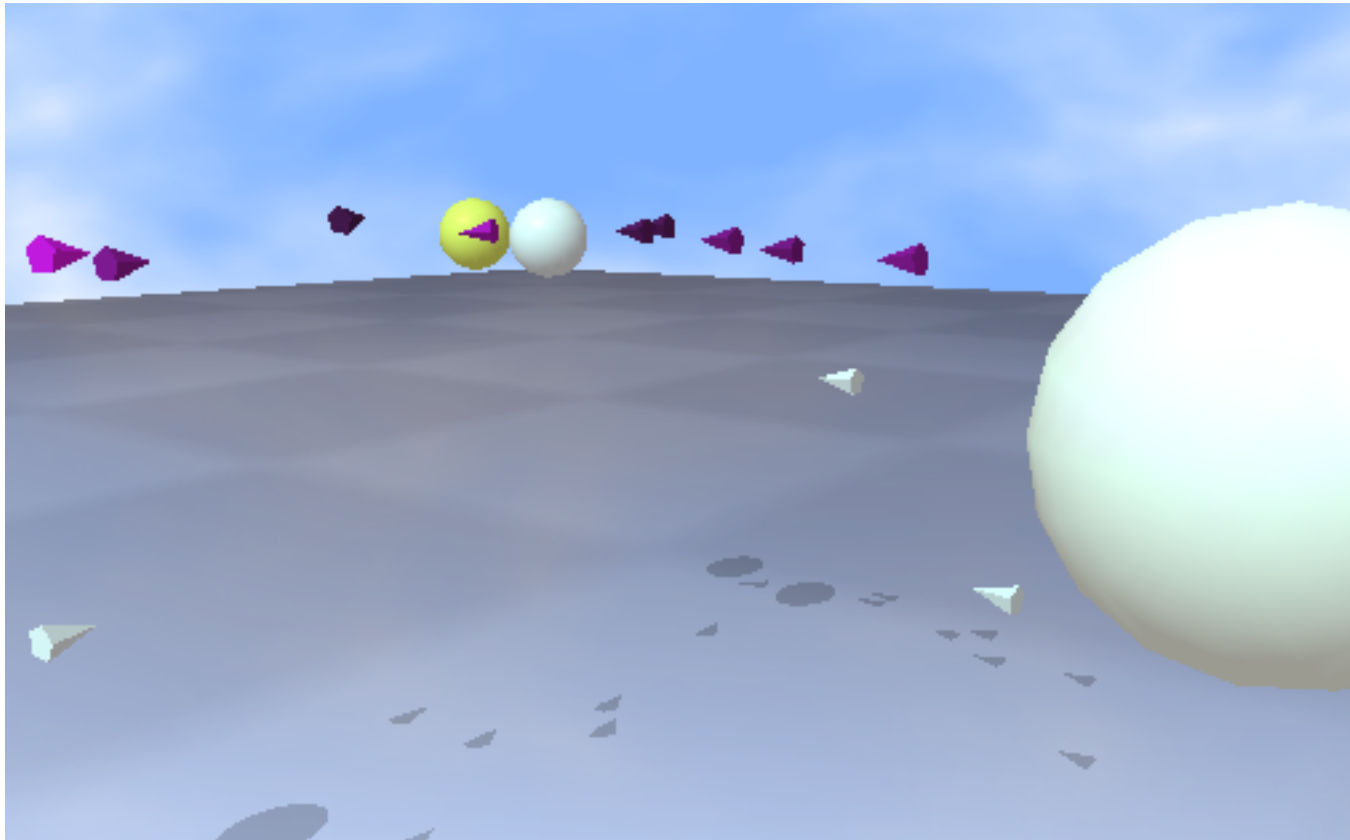


# Automatic Programming of Agents via Multi-type, Self-Adaptive Genetic Programming

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# Overview

Approach and Critical Elements

Technologies:

Push, PushGP, Pushpop, Breve, **SwarmEvolve**

Recent Results:

**Emergence of collective/multicellular organization**

**Environmental/genetic stability and adaptation**

Push/Breve integration v. 0.1

Demo:

**SwarmEvolve 1.5**

**Next Steps**

# Approach & Critical Elements

**Design Approach:** Self-adaptive, multi-type genetic programming for automated or semi-automated agent design.

**Critical Elements:** Autonomy, coordination, **adaptation**, control, evolution.

**Problems:** Can agents be automatically generated for complex, dynamic environments? Can agents evolve to become more adaptable to changing environments?

**Metrics:** Wait time, event response delay, agent lifetime, code parsimony/diversity, evolutionary computational effort, **task completion**.

**Toolkit:** Push programming language for evolved agent programs, PushGP genetic programming system, Pushpop autoconstructive evolution system.

# The Push Programming Language for Evolutionary Computation

Goal: Scale up GP/agents techniques for human-competitive performance in complex, dynamic environments.

Evolve agents that may use:

- multiple data types
- subroutines (any architecture)
- recursion
- evolved control structures
- evolved evolutionary mechanisms

Push supports all of this using simple, mostly standard GP techniques.

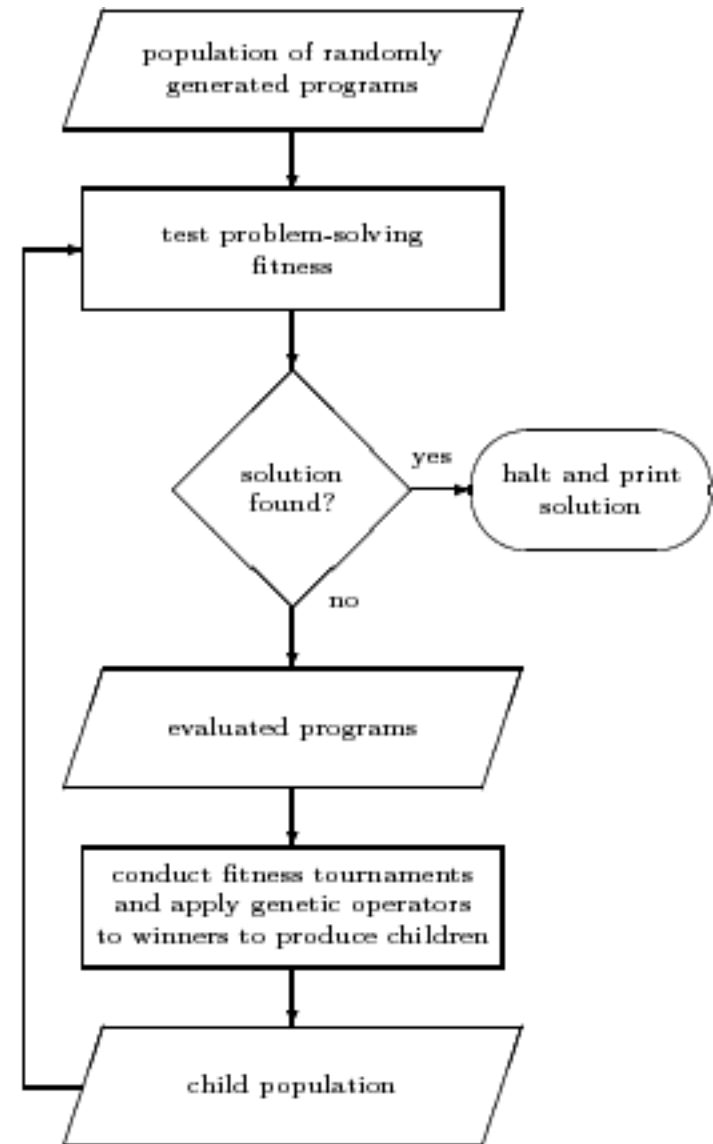
Stack-based language with one stack per type; types include integer, float, Boolean, **code**, child, type, name.

# PushGP

Evolves Push programs using (mostly) standard GP.

Multiple types handled without syntactic constraints.

Evolves modules and control structures automatically.

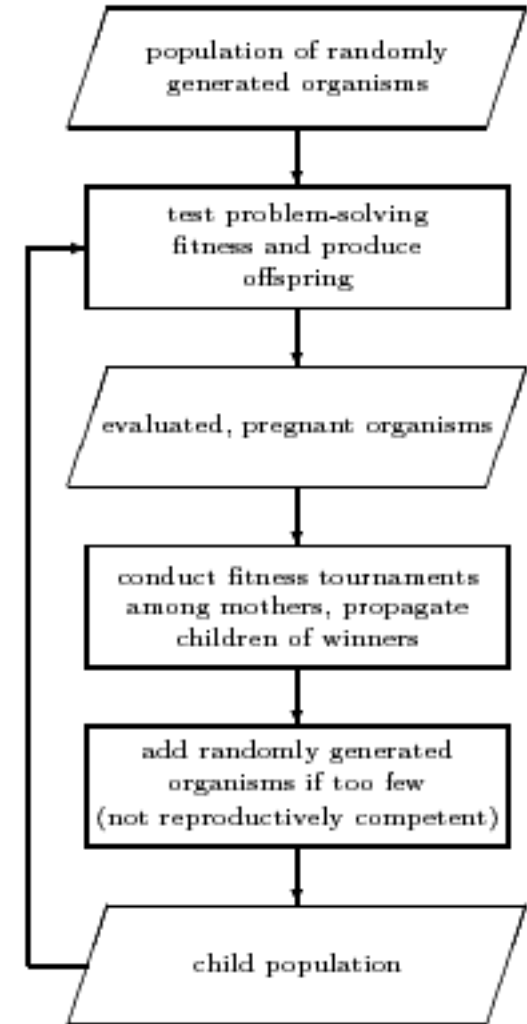


# Autoconstructive Evolution: Pushpop

Individuals make their own children.

The machinery of reproduction and diversification (and thereby the machinery of evolution) evolves.

Radical self-adaptation.



# **Breve: a 3D Environment for the Simulation of Decentralized Systems and Artificial Life**

Written by Jon Klein, <http://www.spiderland.org/breve>

Simplifies the rapid construction of complex 3D simulations.

Object-oriented scripting language with rich pre-defined class hierarchy.

OpenGL 3D graphics with lighting, shadows, and reflection.

Rigid body simulation, collision detection/response, articulated body simulation.

Runge-Kutta 4th order integrator or Runge-Kutta-Fehlman integrator with adaptive step-size control.

# Breve Swarm

by Jon Klein, after Craig Reynolds

$$\begin{aligned} \text{acceleration} = & p_1^* [\text{away from crowding others vector}] \\ & + p_2^* [\text{towards world center vector}] \\ & + p_3^* [\text{average neighbor velocity vector}] \\ & + p_4^* [\text{towards neighbor center vector}] \\ & + p_5^* [\text{random vector}] \end{aligned}$$



# SwarmEvolve

On-Line evolution of goal-directed swarms

Multiple species

$\rho_6^*$  [away from crowding other species vector]

Randomly moving energy sources:

$\rho_7^*$  [towards closest energy source vector].

Energy costs:

- Colliding with one another
- Being outnumbered (by species) in neighborhood
- Giving birth
- Surviving (per simulation cycle)

Upon death (energy = 0), parameters replaced with mutated version of fittest of species

Fitness metric = age \* energy

# SwarmEvolve 1.5

- Food consumption/growth
- Birth near mothers
- Corpses
- Food sensor, inverse square signal strength
- GUI controls and metrics
- Feeders redesigned, increased in number
- OEF correspondence increasing

[\[view movie\]](#)

# Emergence of collective/multicellular organization

Observed behavior: a cloud of agents hovers around an energy source. Only the central agents feed, while the others are continually dying and being reborn.

Can be viewed as a form of emergent collective organization or multicellularity.

Facilitated by “birth at death location” implementation.

To appear in proceedings of *Beyond Fitness: Visualising Evolution*, a workshop at ALife 8.

[\[view movie\]](#)

# Environmental/Genetic Stability and Adaptation

Food supply as a function of environmental stability and mutation rate:

		MUTATION		
		low	med	high
STABILITY	low	54%	17%	18%
	med	43%	12%	10%
	high	55%	14%	12%

Preliminary data (2 runs/condition) averaged over first 10,000 time steps of each run.

**[Demo: SwarmEvolve 1.5]**

# Next Steps

Enhance complexity/realism/OEF integration.

Species-specific controls and metrics.

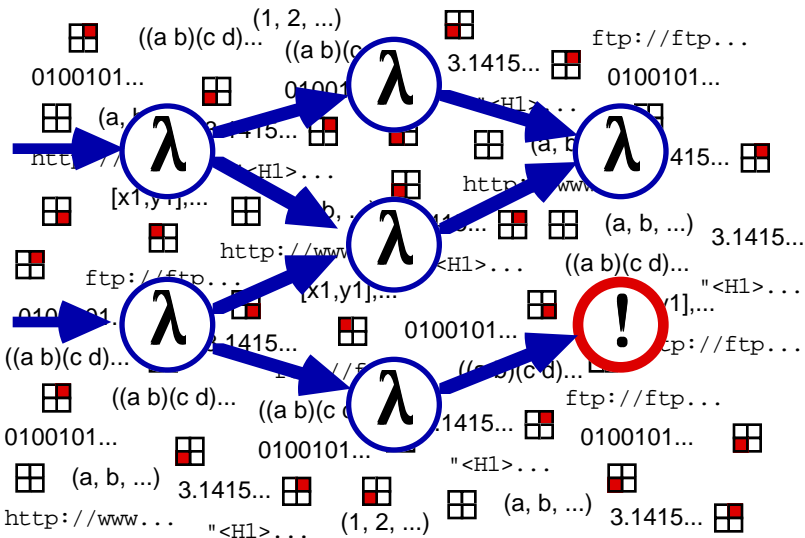
Structured feeder behavior; agent-responsive.

Leverage Push/Breve integration for evolution of arbitrary agent control programs and group (species) distinctions.

Integrate MIT/BBN elementary adaptive modules.

Provide “evolution” components for Taskable Agent Software Kit.

# Multi-Type, Self-Adaptive Genetic Programming for Complex Applications



## New Ideas

- Richly heterogeneous data can be flexibly integrated in programs produced by stack-based genetic programming.
- Explicit code manipulation allows for automatic emergence of modules and evolved program architecture.
- Self-adaptive construction of evolutionary mechanisms enhances fit to problem environments.

## Impact

- Evolved agents for heterogeneous, dynamic environments.
- Broader range of applications for automatic programming technologies.
- Automatic programming with less configuration by users.

## Schedule

