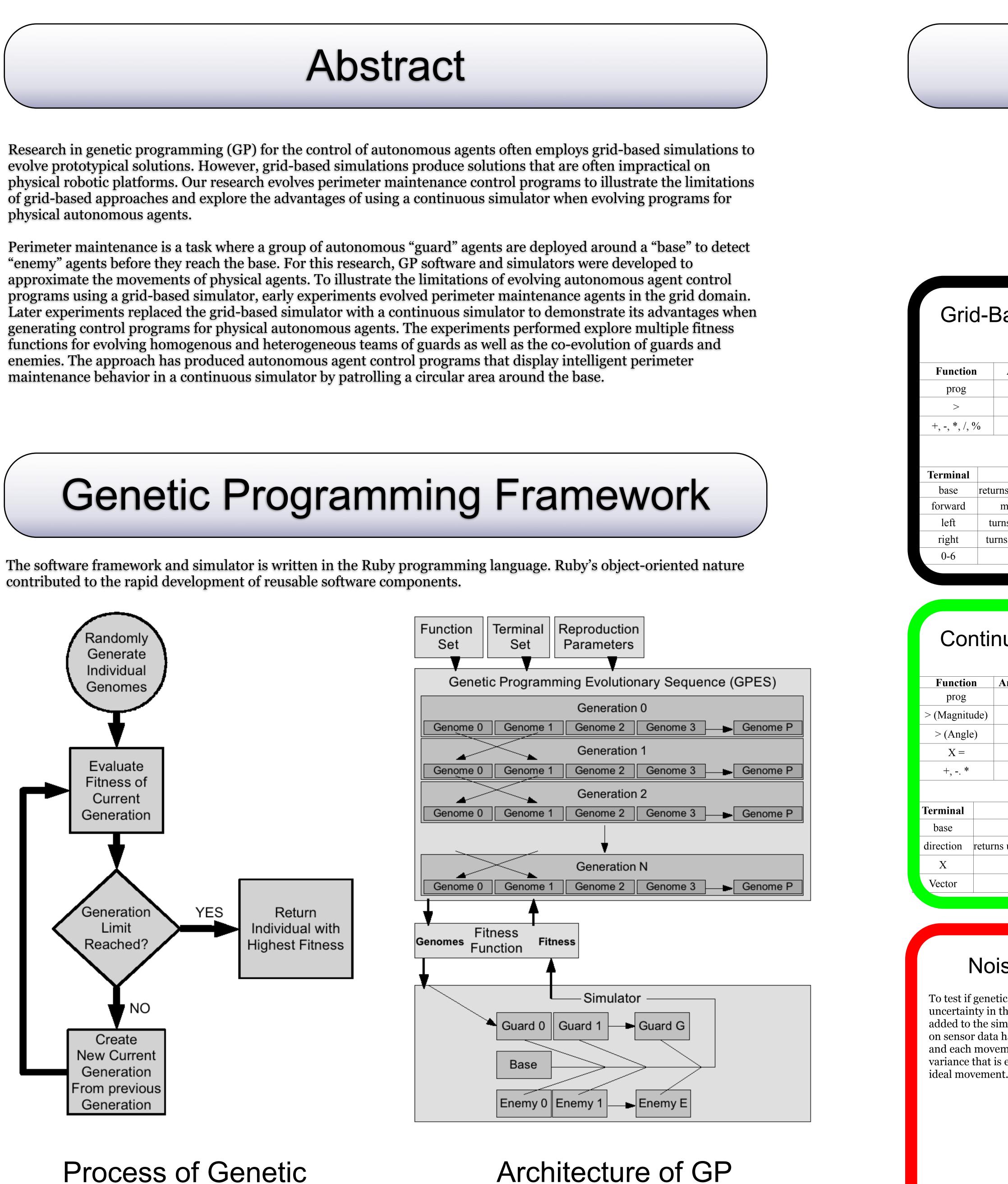
By: Scott O'Dell

physical autonomous agents.

contributed to the rapid development of reusable software components.



Programming

Genetic Programming of Autonomous Agents Bradley University 2011

Framework

Advisor: Dr. Joel Schipper

Results of Evolutionary Sequences

Evolution of Guards

Each guard is controlled using the same program. Enemies start on the edge of the simulation and moves directly toward the base.

Grid-Based Simulations

Function Set

Arity	Pseudo-code
2	(a) then return(b)
4	if $(a > b)$ then (c) else (d)
2	[standard integer arithmetic]

Terminal Set

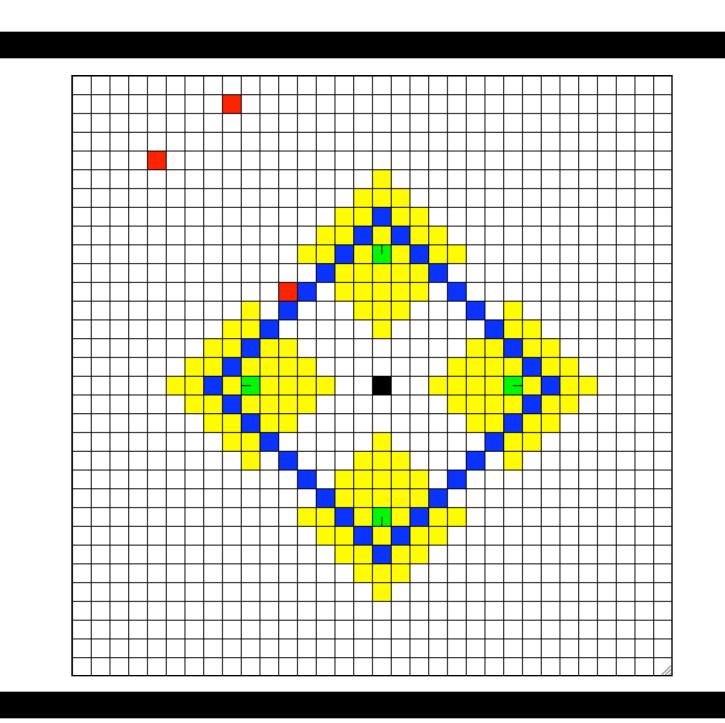
Effect urns Manhattan distance from guard to base moves agent forward, returns "base" turns agent left 90 degrees, returns "base" turns agent right 90 degrees, returns "base" constant integers

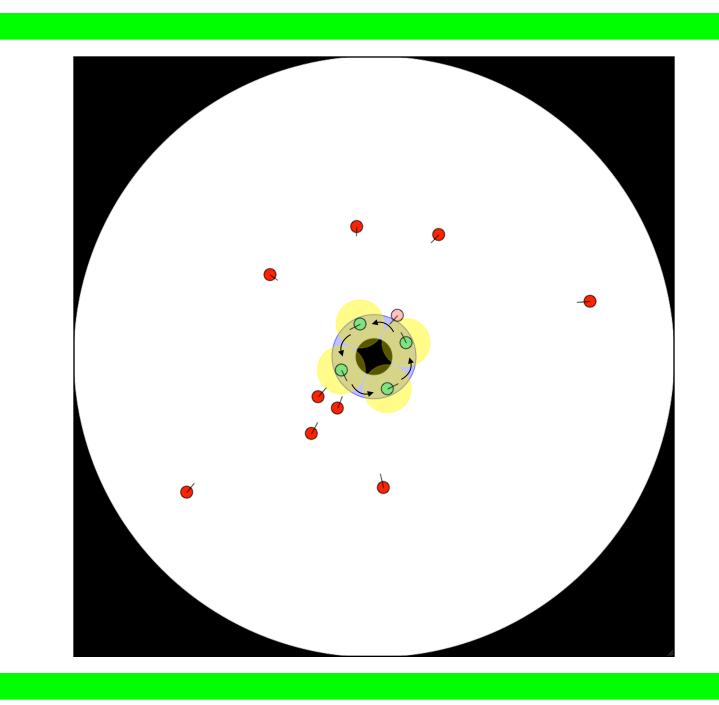
Continuous Simulations

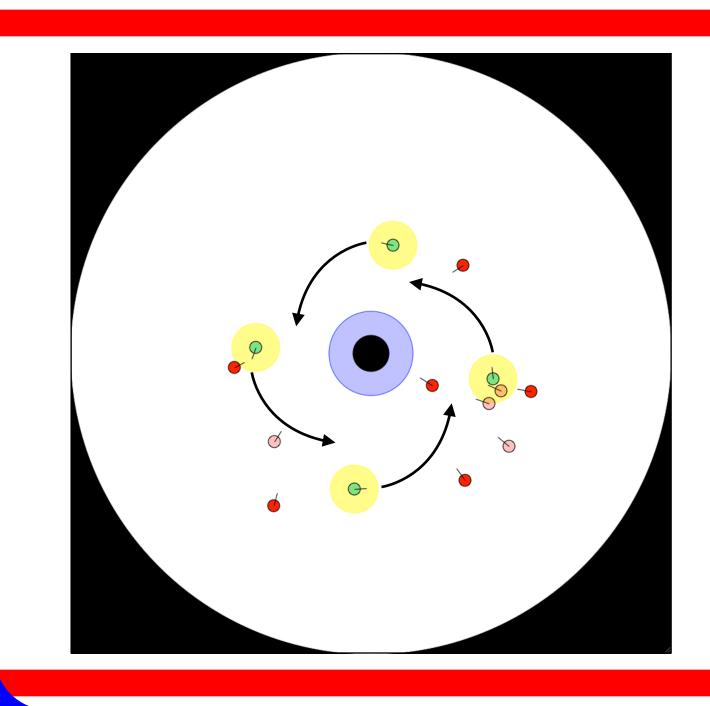
rity	Pseudo-code	
2	(a) then return (b)	
4	if (a.mag > b.mag) then (c) else (d)	
4	if (a.ang > b.ang) then (c) else (d)	
1	variable $X = (a)$	
2	[standard vector arithmetic]	
Terminal Set		
Effect		
returns vector from guard to base		
unit vector representing guard's heading		
	return (variable X)	
[static vector]		

Noisy Simulations

To test if genetic programming can deal with uncertainty in the environment, Gaussian noise was added to the simulations. Any terminal that is based on sensor data has noise with a constant variance. and each movement is subject to noise with a variance that is equal to 1/10th the magnitude of the









Co-evolution of Guards & Enemies

Enemies are evolved with the same primitive set as the guards with the addition of terminals that provide the position of the closest guard. Evolving the enemies simultaneously with the guards produces less exploitable solutions from the guards.

