
Experiments on the Automatic Evolution of Protocols using Genetic Programming

Lidia Yamamoto and Christian Tschudin
Computer Networks Group, Computer Science Department
University of Basel, Switzerland
<http://cn.cs.unibas.ch/>

2nd IFIP International Workshop on
Autonomic Communication (WAC 2005)
Athens, Greece, October 3-5, 2005

Outline

- AC Self-management
- Adaptation and evolution
- Evolving autonomic network software
- Framework for protocol evolution
- Experiments
- Conclusions, next steps

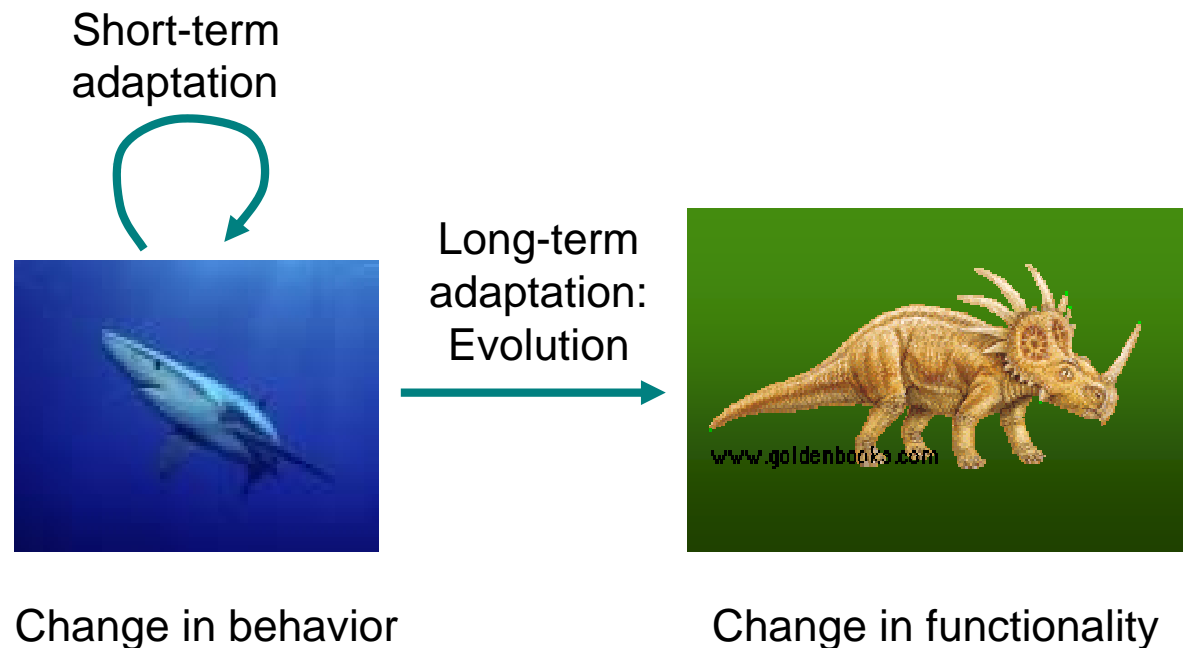
Context

- BIONETS: Biologically-inspired Networks and Services
 - Situated and Autonomic Communication FET Initiative
 - 4-year project starting January 2006
 - Central question: How to make protocols and services evolve automatically during usage: runtime evolution
 - 1st step: Parameter evolution: analogy: genetic algorithms
 - 2nd step: Code evolution: analogy: genetic programming
- This talk:
 - Code evolution experiments using genetic programming

Introduction

- Autonomic Communication (AC)
 - Network elements conspire to do what we want
 - Self-*, including self-management
- **Fully self-managing networks** networks must take complete care of themselves, including **software maintenance**
 - Detect and correct software failures
 - Optimize software for specific context
 - Constantly keep on target and improve itself, without direct human intervention: autonomic!
- Thus: full AC requires **automated software evolution**
 - Otherwise humans must intervene to modify software

Adaptation and Evolution



Evolving Autonomic Network Software

- Automated **evolution**
 - Functional scaling: beyond self-adaptive software
- Automatically generate new, better replacement code
 - ⇒ **Self-modifying code**
- During operation: **runtime evolution**
 - Resilience and survivability
 - Potentially hostile operational environment
 - Heterogeneous networks and users, competing interests
 - Code errors, malicious code, non-trusted parties, ...
 - Fully cooperative learning scheme not realistic
- A possible path: **distributed on-line genetic programming**

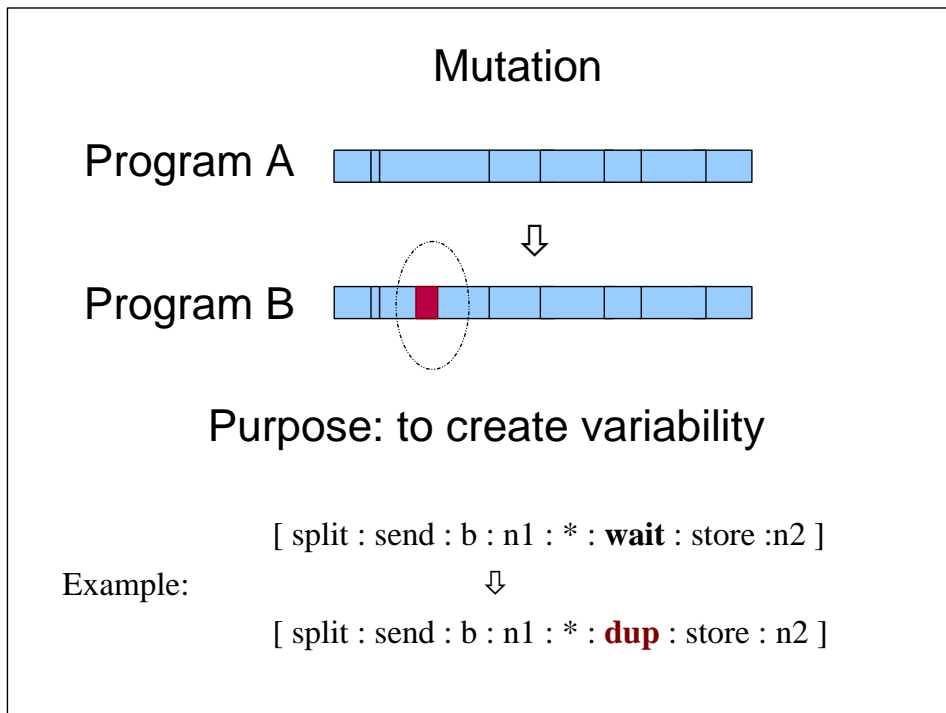
Evolving Autonomic Network Software

- Genetic Programming
 - Machine learning method for synthesizing programs
 - Agnostic program transformations: crossover, mutation
 - Natural selection: survival of the fittest
 - In general off-line: optimal solution as output
- **Distributed on-line genetic programming**
 - Program transformations rely on code mobility
 - Non-disruptive execution of synthesized programs
 - Competitive/hostile environment
 - Natural selection pressure ⇔ Survivability
 - Redundancy ⇔ Resilience

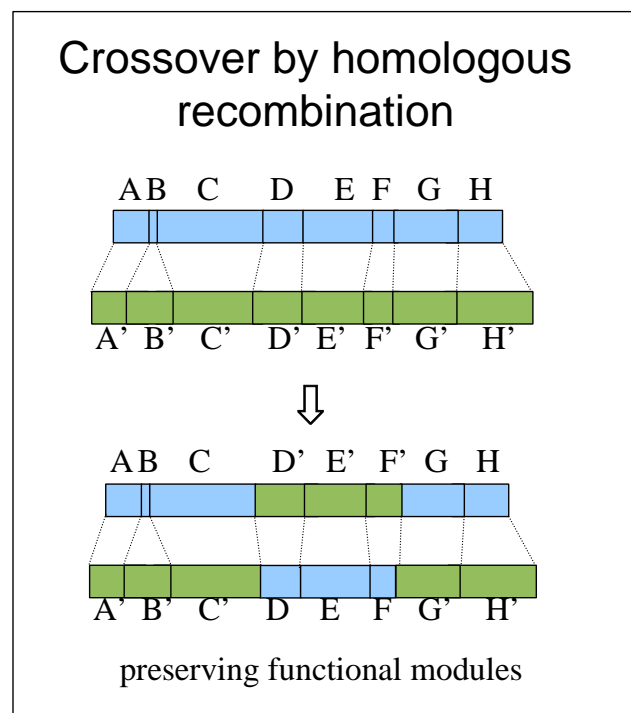
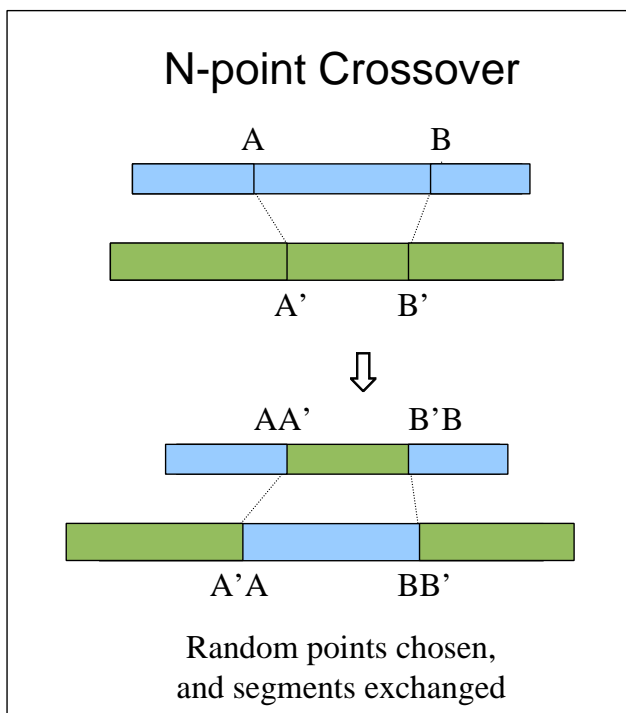
Framework for Protocol Evolution

- Start with working protocol implementations
- Evolution via genetic programming
- **Fraglets** chemical programming model
 - Easy GP: any code fragment (fraglet) is a valid program
 - Parallelism ⇔ redundancy ⇔ resilience
 - Resilience to code loss: initial results in WAC 2004
- Still off-line (simulations), but assumptions for on-line
 - Small population
 - Limited number of generations
 - Resilience: minimize service disruptions due to malfunctioning code

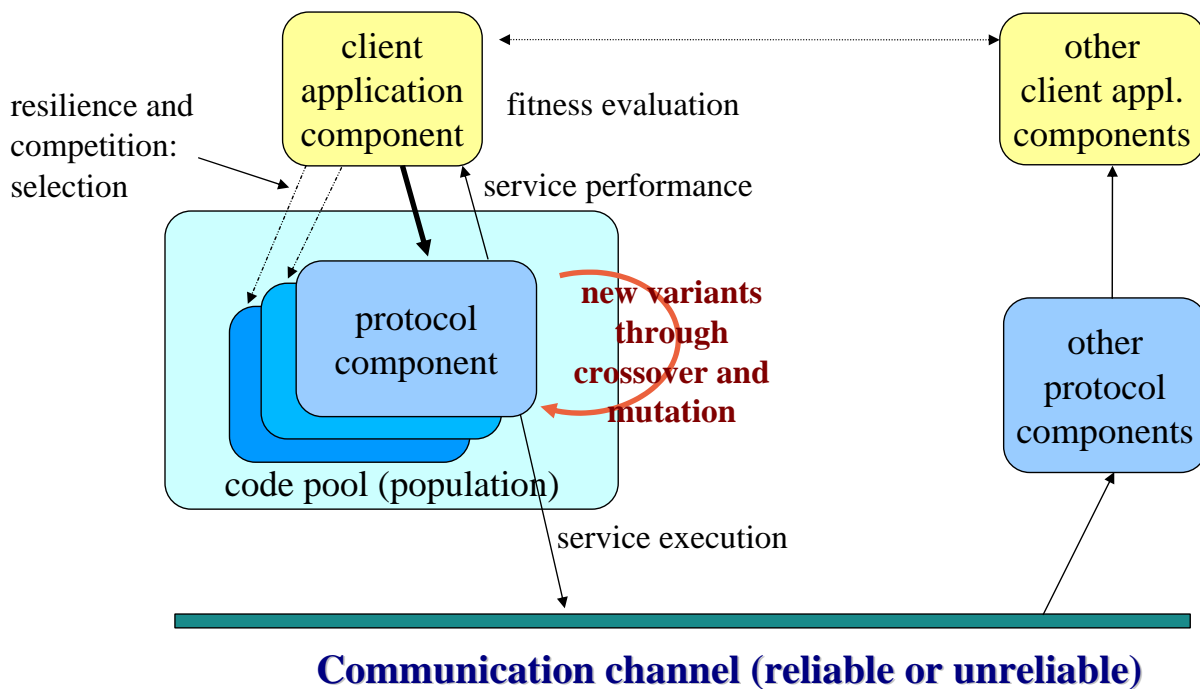
Genetic Operators



Genetic Operators



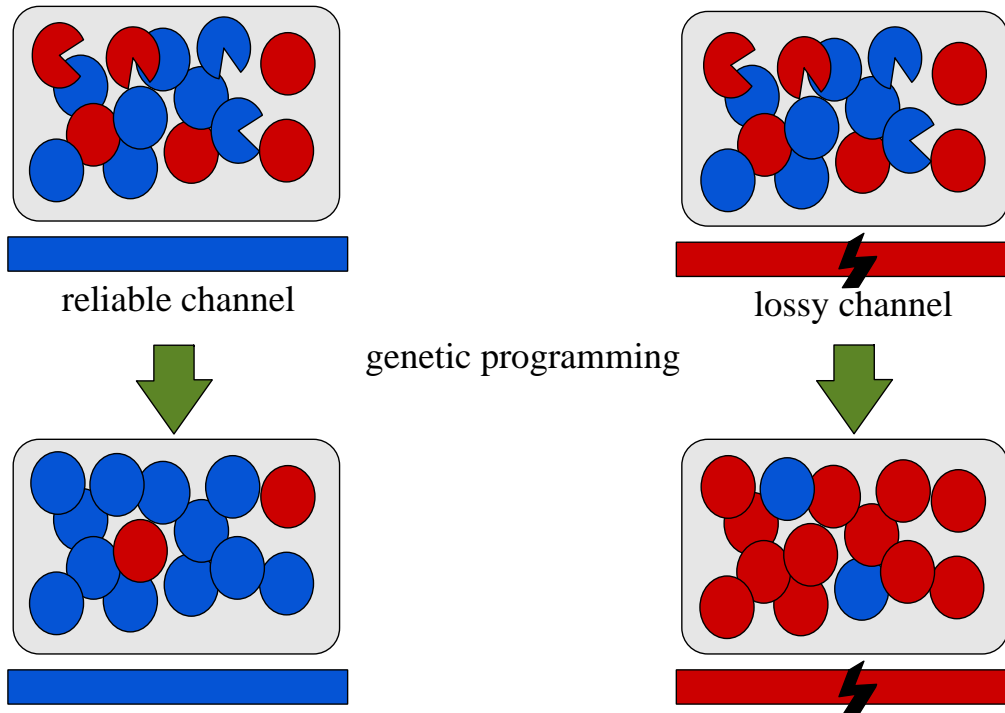
Framework for Protocol Evolution



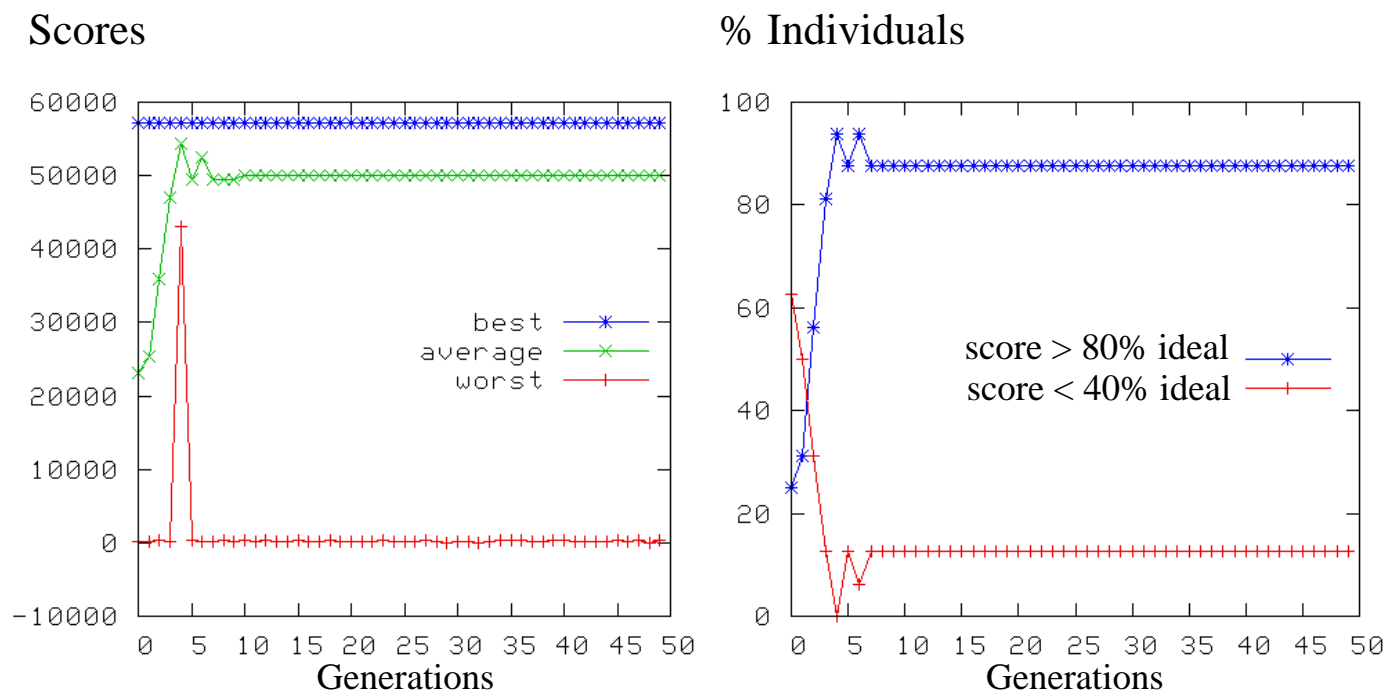
Experiments

- Reliable transmission protocols
 - Initial population: Multiple variants (alternative implementations): more/less efficient
- Two types of channel
 - **Reliable** transmission channel (no packet loss)
 - **Unreliable** (lossy) channel
- Application: requires 100% reliable transmission
 - Underlying protocol must retransmit lost packets if channel is unreliable
 - Fitness evaluation = measure of reliability = score
- Adaptation: selection and dissemination of suitable code

Adaptation Experiment

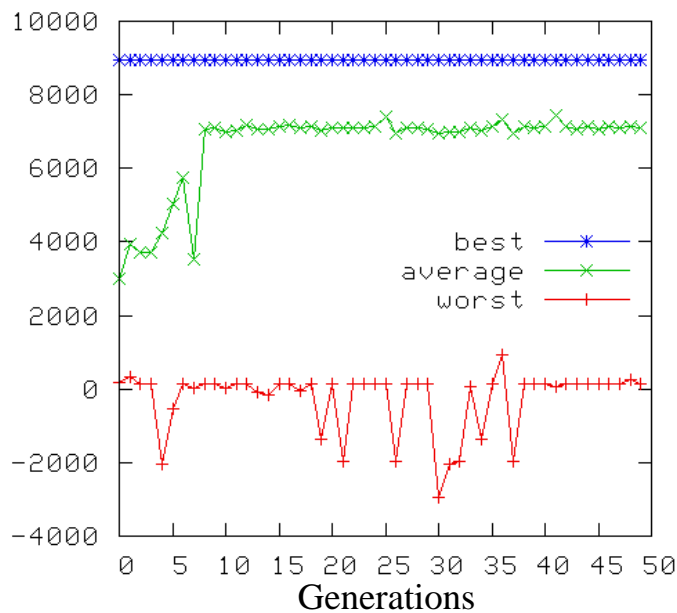


Adaptation Experiment: No Packet Loss

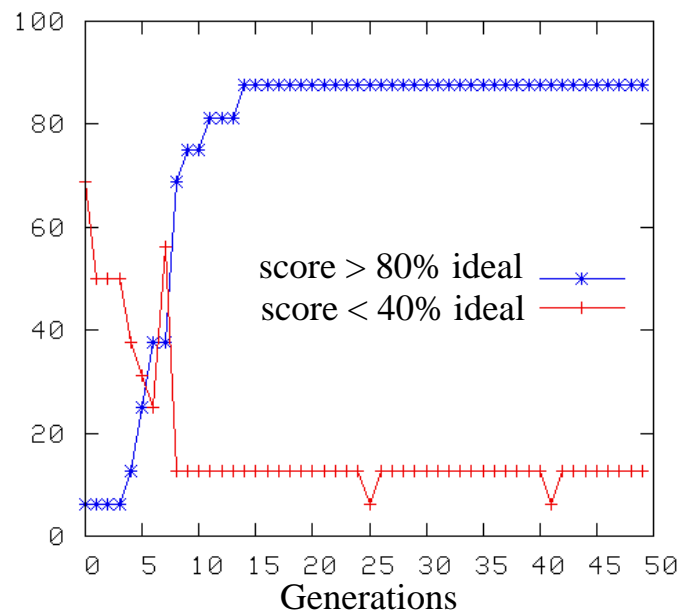


Adaptation Experiment: With Packet Loss

Scores



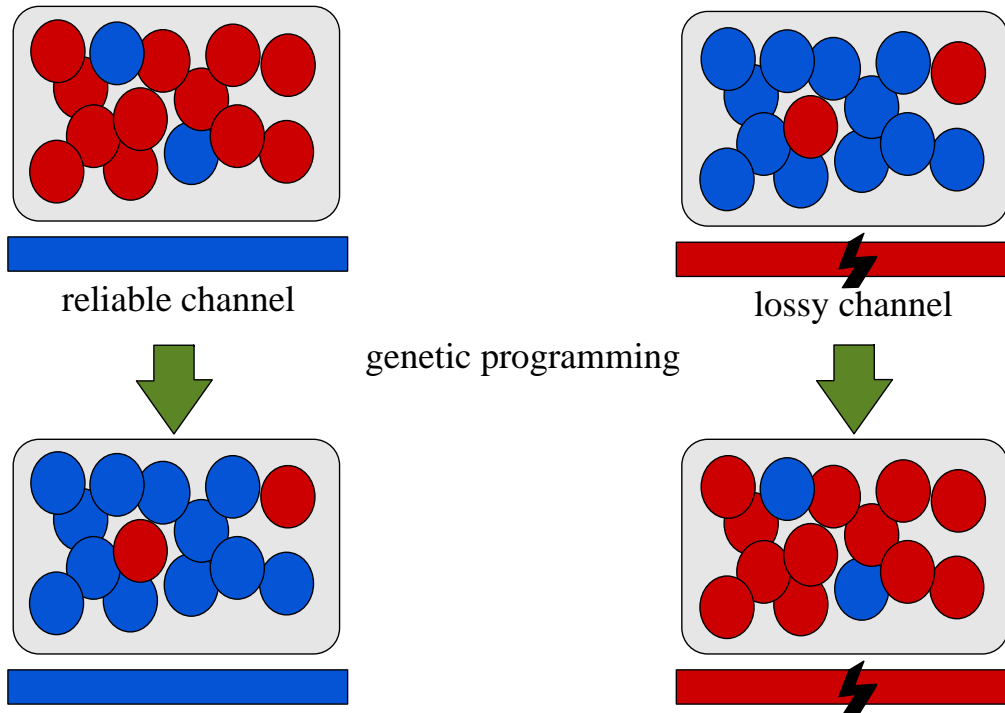
% Individuals



Adaptation Experiment

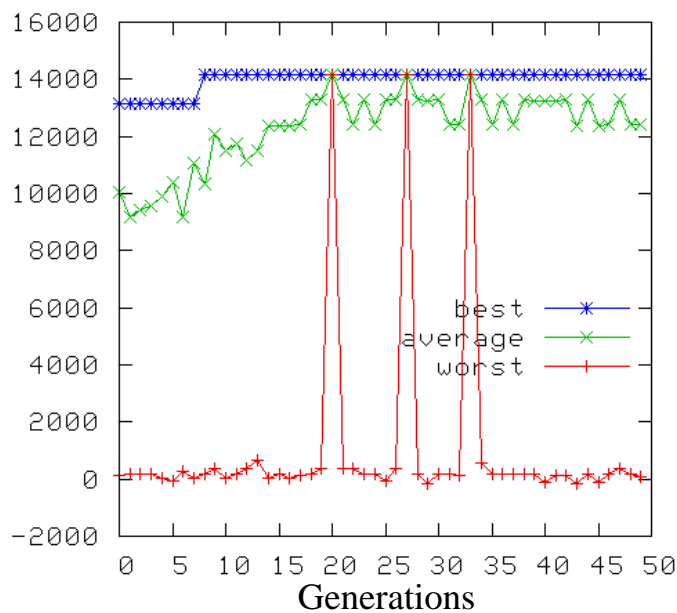
- Successful individuals (ideal scores) remain and also spread in the population
- After a few generations
 - More than 80% of individuals in the population have a score close to the ideal
 - Less than 20% of individuals have poor score
- But genetic variability is severely reduced, population becomes very uniform

Re-adaptation Experiment

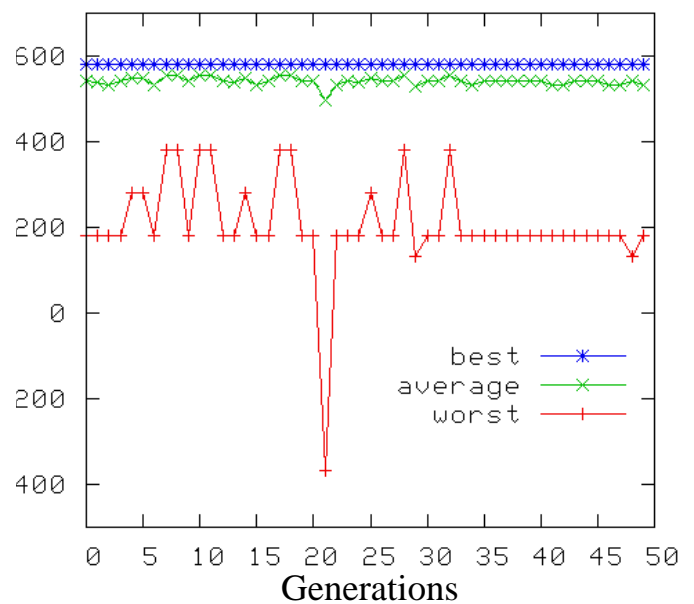


Re-adaptation Experiment (1)

Scores lossy to non-lossy

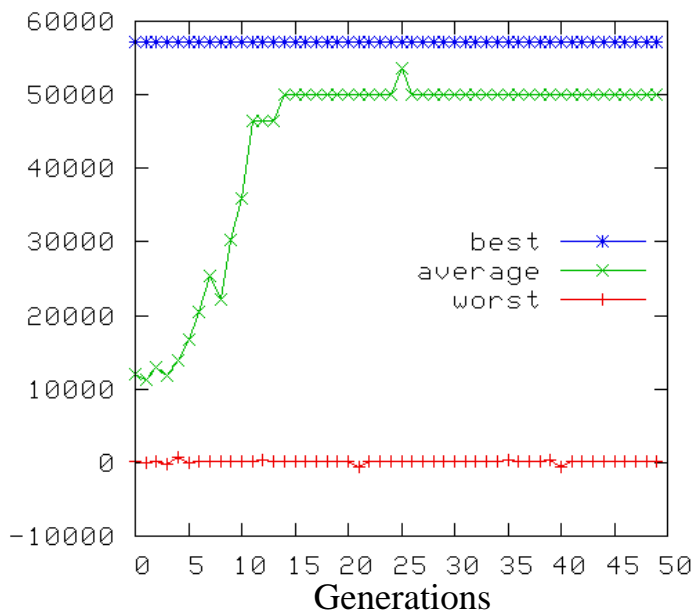


Scores non-lossy to lossy

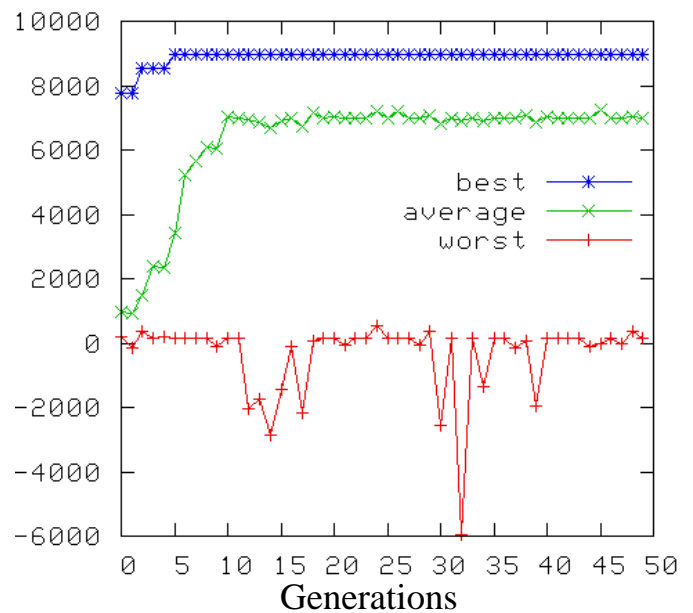


Re-adaptation Experiment (2)

Scores lossy to non-lossy



Scores non-lossy to lossy



Is On-Line Protocol Evolution Possible?

➤ Answers so far:

- Can "survival of the fittest" strategy really make best protocols spread in the population?
 - Yes, and quickly, but then genetic variability is lost
- Can they readapt?
 - Yes, provided that at least one already adapted individual is present in population

Lessons Learned

- Code modification via genetic operators:
 - Crossover
 - Homologous recombination is safe but limited
 - Unable to "create" really new code
 - Unbounded crossover (random points) leads to "intron growth" phenomenon
 - Rise of polluted code containing useless garbage
 - Mutation
 - Currently too slow and random
 - Low probability to produce viable individuals

Conclusions

- Experiments show conditions under which adaptation and re-adaptation are possible: first steps towards evolution
- Resilience at the population level achieved via selection of best and elimination of unsuitable code variants
- GP can do much better than random search, but size of search space still too vast ($\sim 10^{200}$ for these simple experiments)

Next Steps

- Improve genetic operators
 - Compromise between code safety and variability
 - Hybrid operators using deterministic and formal methods
- Resilience at the individual level
 - Inspired by genome redundancy, metabolic pathways
- Decentralized fitness evaluation
 - Redundant protocol execution circuits
 - Trust and reputation
- Propagation of evolved protocols
 - User and node mobility, code mobility