


*Genetic Programming  
of  
Process Decomposition  
Strategies  
for Evolvable Hardware*



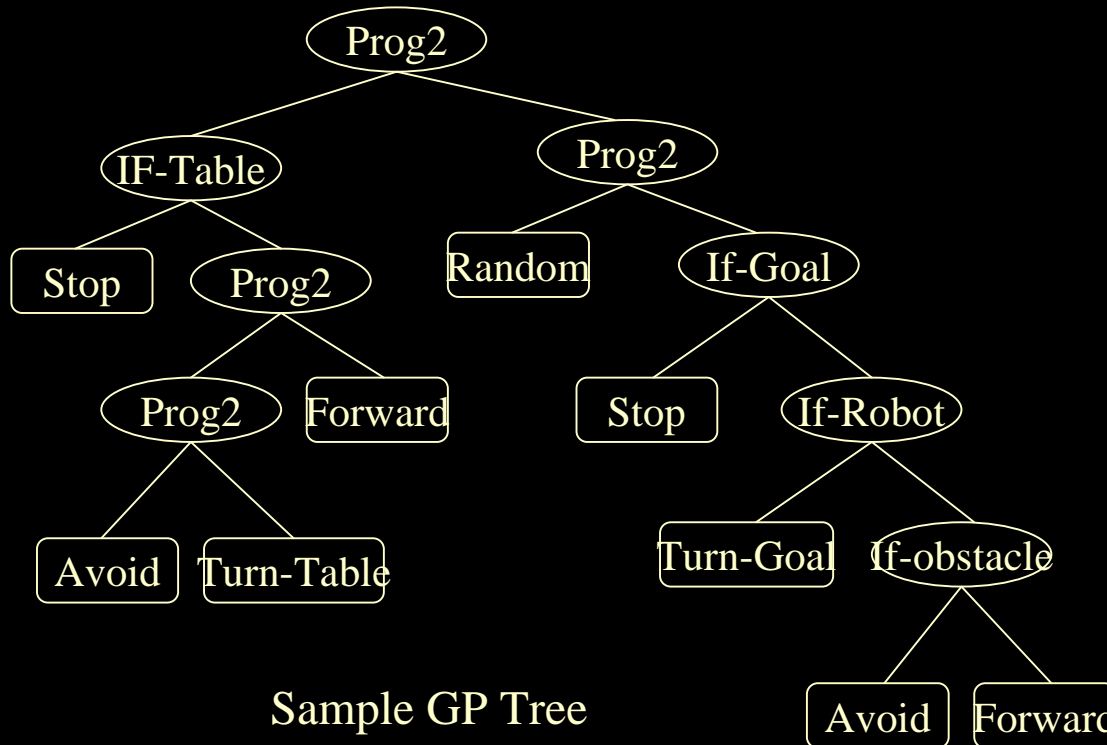
# *Outline*



- Genetic Programming (GP)
- Evolvable Hardware (EH)
- Hardware Evolution of GP Trees
- Context Switching
- Implementation and Experiments
- Conclusion and Future Work

# Genetic Programming (GP)

This is a program for  
a mobile robot



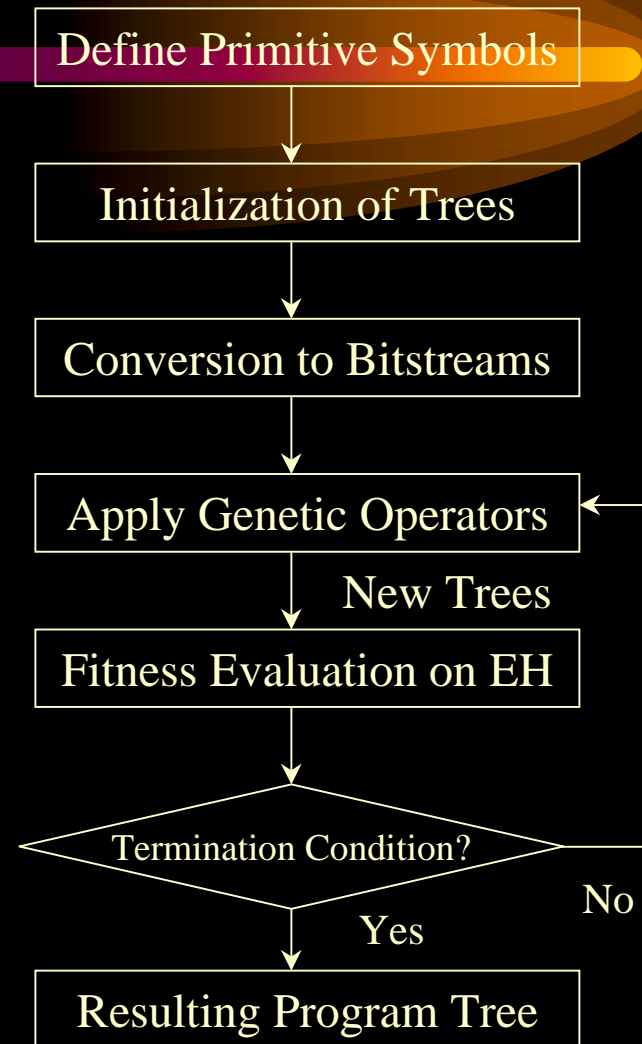
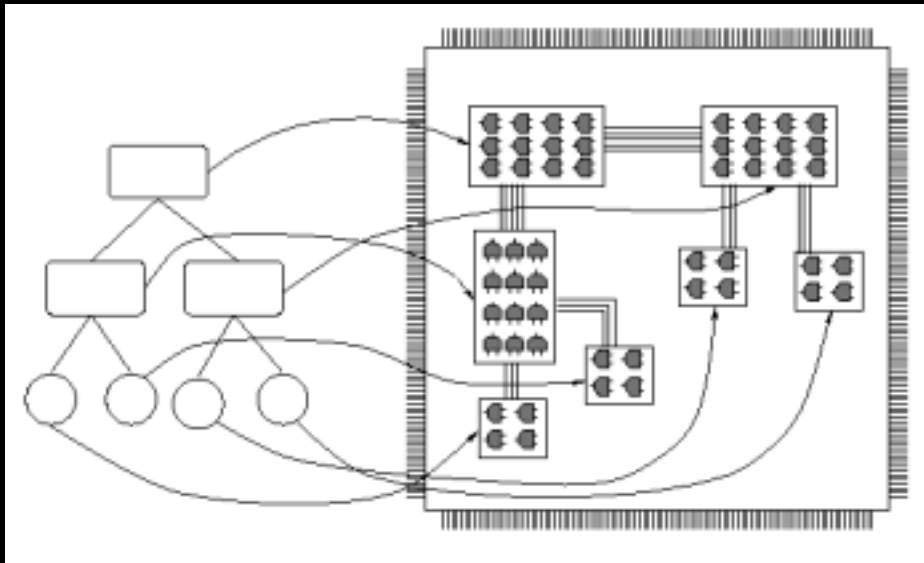
- + Automatic evolution of computer programs
- + Tree-structured chromosomes
- + Expressive power
- + Generality
- + Easy to incorporate prior knowledge
- Time complexity (fitness evaluation)
- Space complexity

# *Evolvable Hardware (EH)*



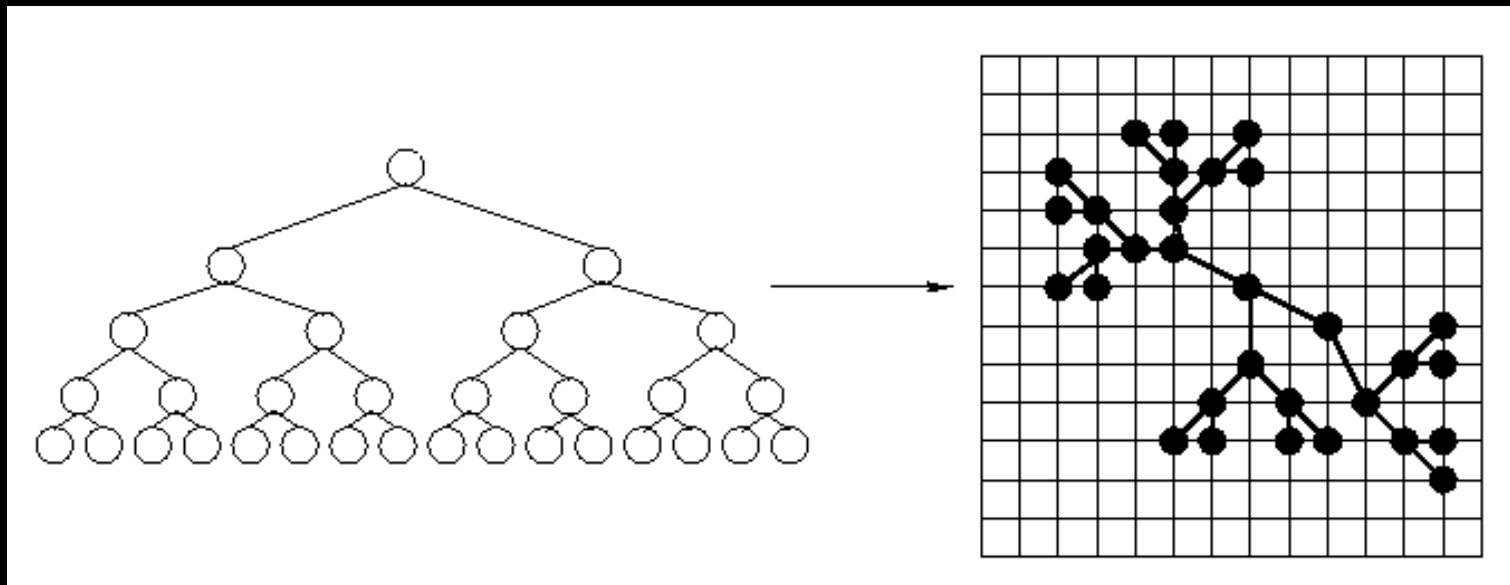
- Run-time reconfigurability
- Higher performance than general-purpose processors
- More flexible than ASICs
- On-line learnability
- Customization

# Hardware Evolution of GP Trees

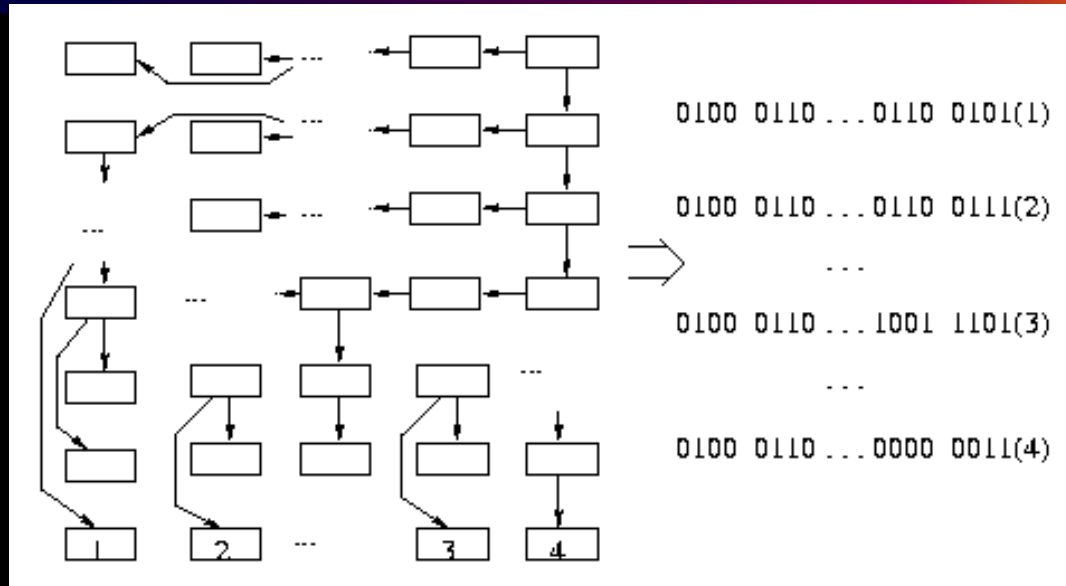


# *Difficulty in Hardware Evolution of GP Trees*

- On-chip representation of tree structures
- Routing problem
- Hardware resource utilization



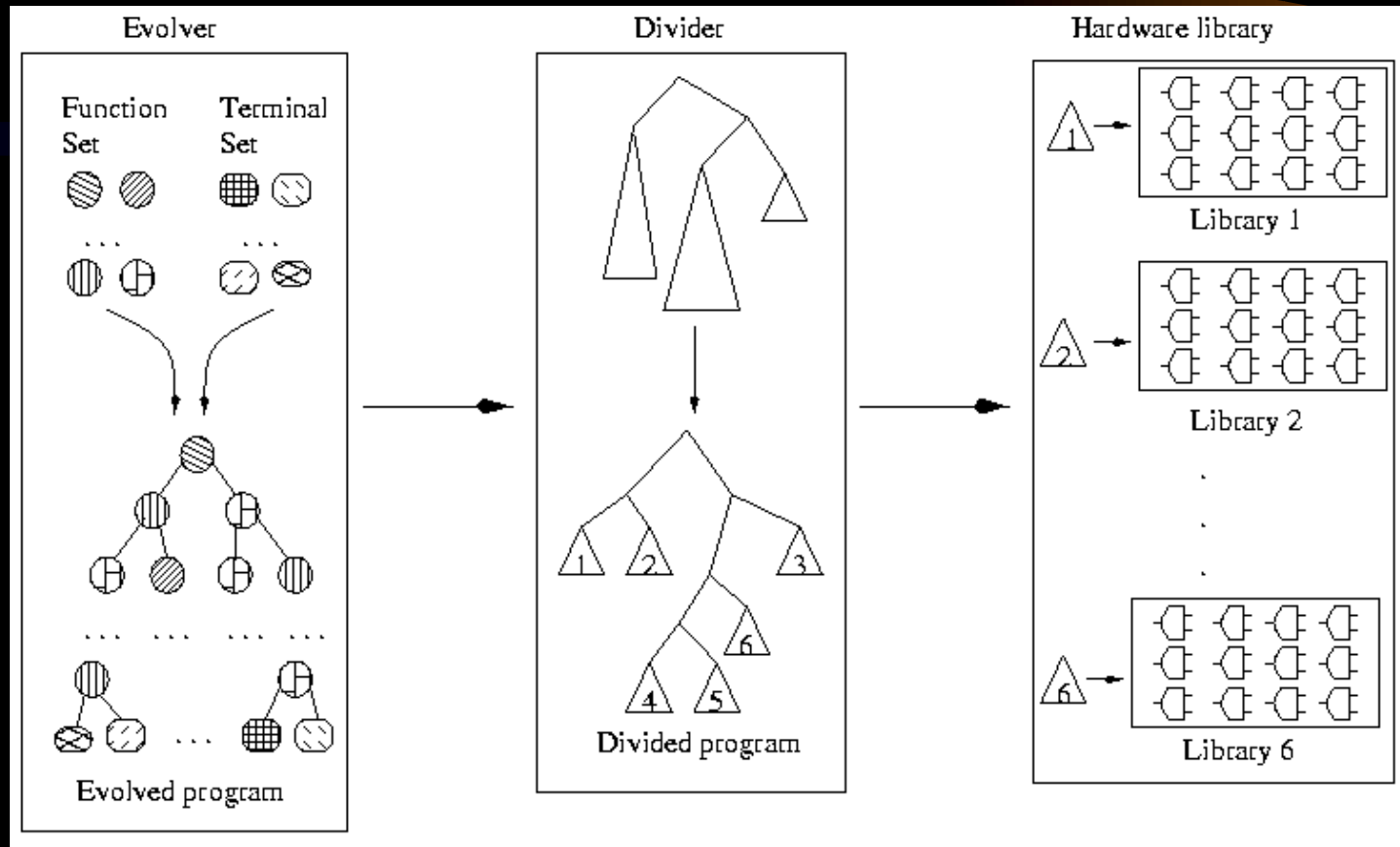
# First Approach: Linear Representation



- Comparison of resource utilization

	Tree Representation	Linear Representation
Total Resource	80.9%	86.7%
Placement Resource	37.8%	49.2%

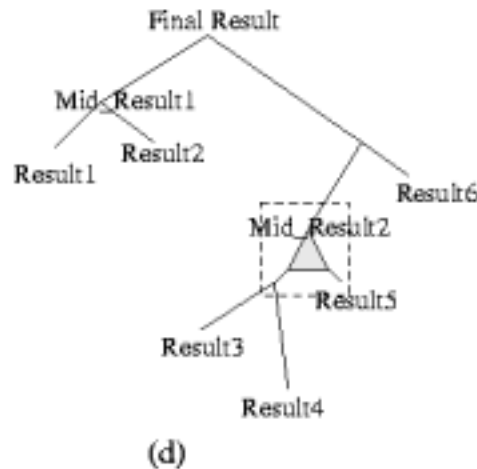
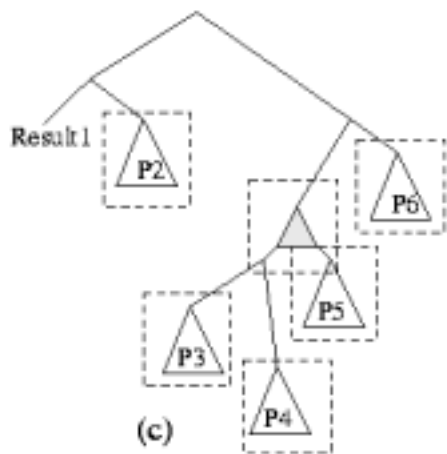
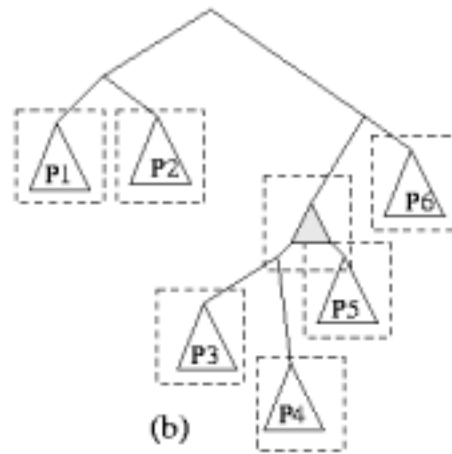
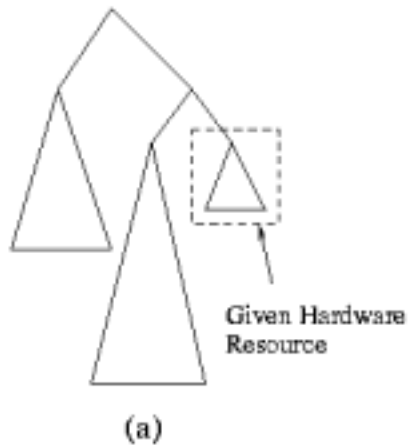
# Second Approach: Context Switching



- Separate implementation of subtrees of a GP tree
- Reduction of redundancy
- Better utilization of chip resources



# Context Switching: Process Decomposition



- A GP Tree is decomposed into a number of sub-trees.
- Based on the resource necessity of sub-trees.
- Number of nodes in the sub-tree

# Context Switching: Hardware Library

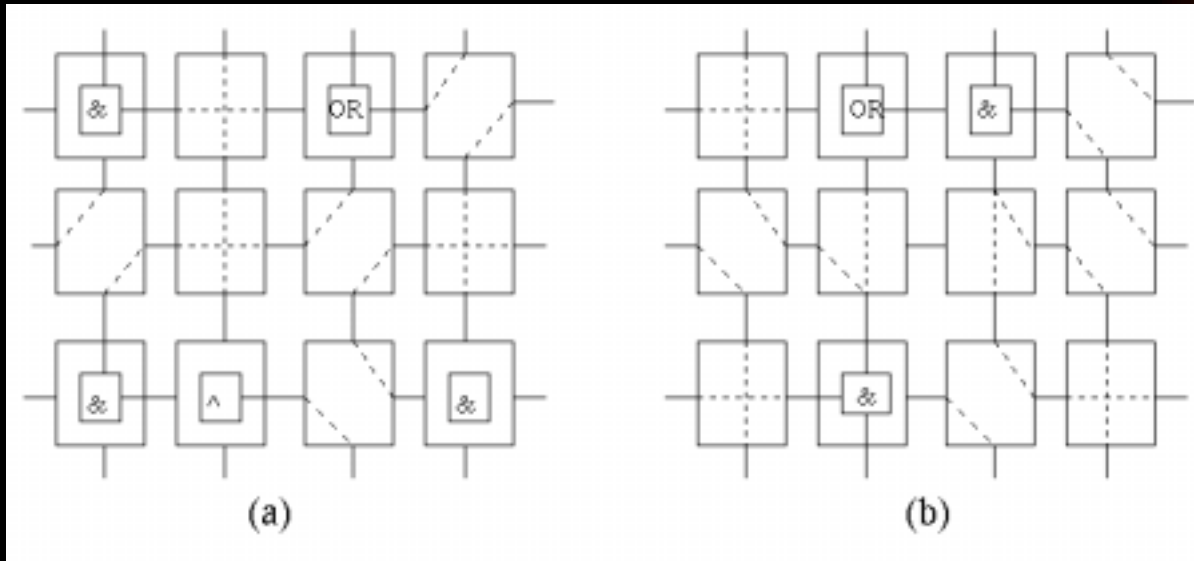
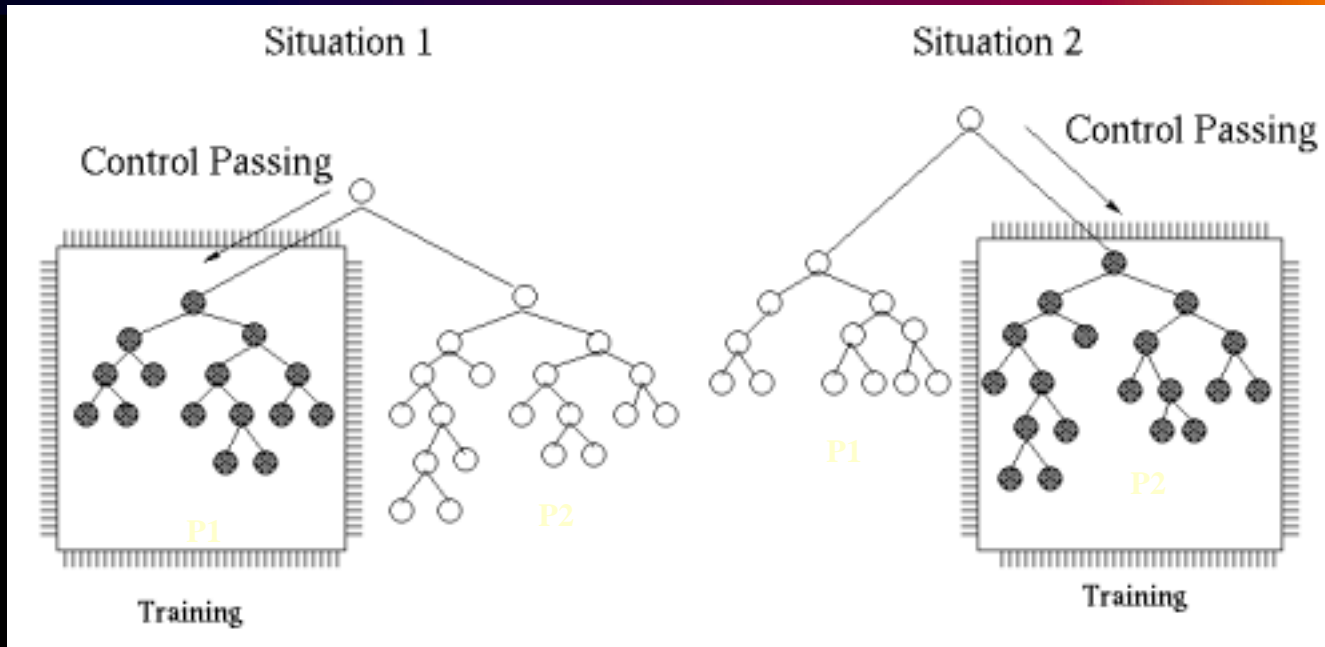


Illustration of hardware library

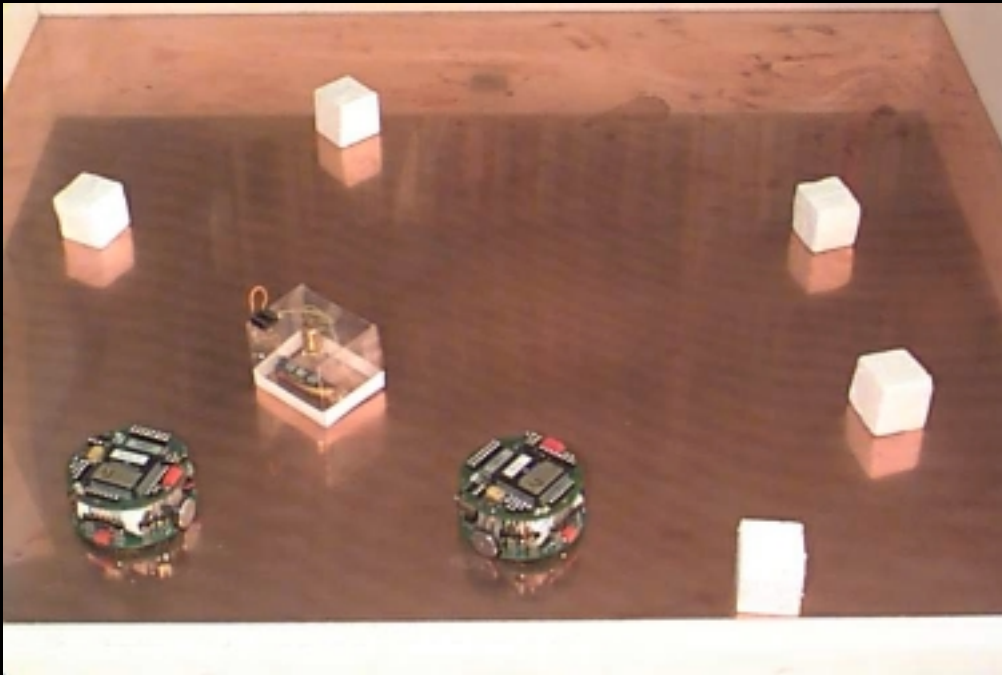
- Library: storage of hardware configuration information for subtrees
- Nodes of a GP tree are interchangeable
- Drawback
  - Possibility of resource waste

# Training of GP Trees with Context Switching



- A sub-tree is chosen depending on the input values.
- Then, this sub-tree is trained by varying and selecting fitter subtrees.

# *Test Bed: Evolving Controllers for Autonomous Robots*



- Transportation of an object to the goal (light)
- Cooperation of two robots
- Khepera
- Xilinx XC6216

# Setup of GP for Robot Control

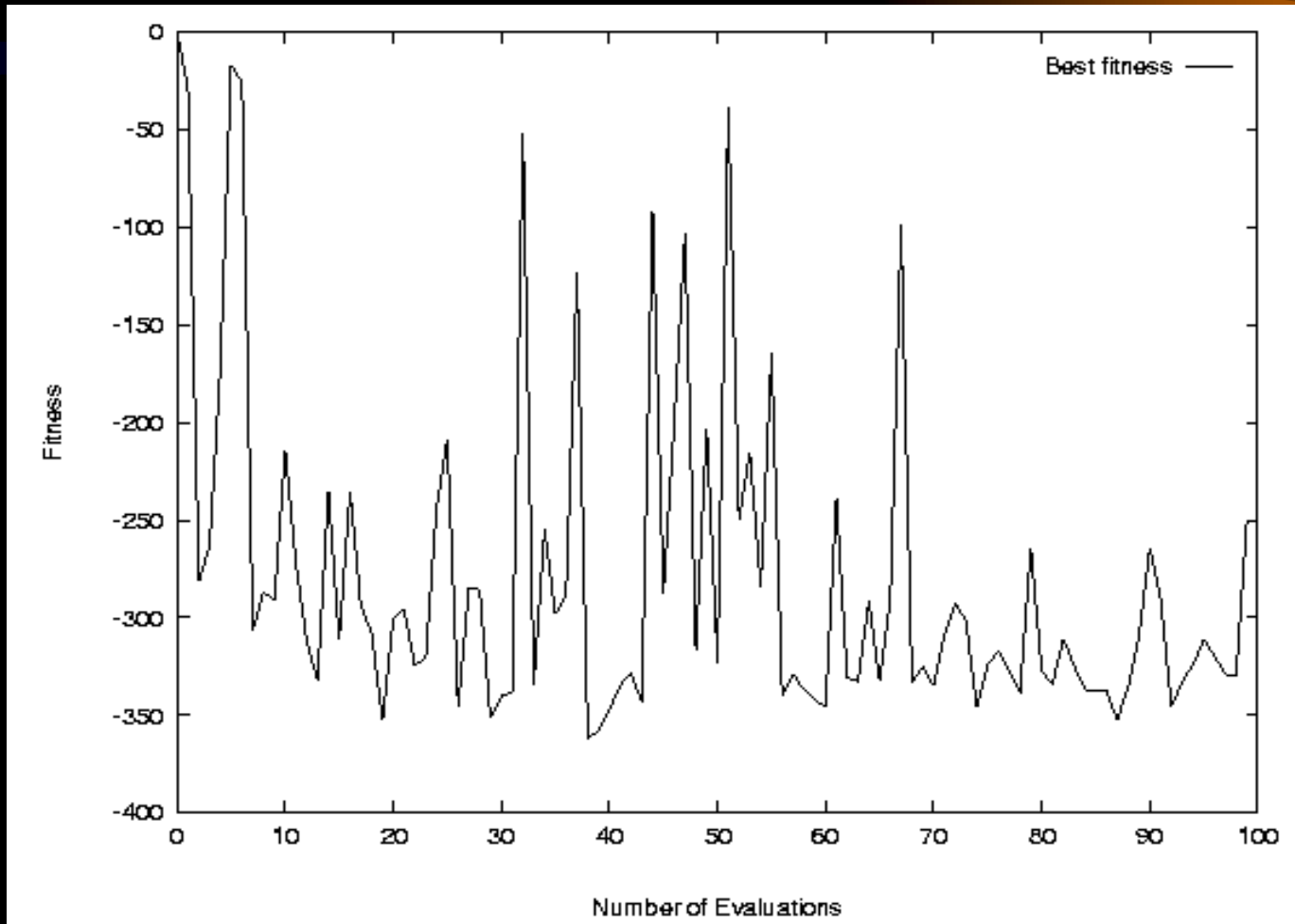
- Function nodes
  - *IF\_OBJ, IF\_GOAL, IF\_FORWARD, IF\_OBS1~4*
- Terminal nodes
  - *MOVE\_FORWARD, MOVE\_FORWARD & TURN\_LEFT, MOVE\_FORWARD & TURN\_RIGHT, MOVE\_BACKWARD, TURN\_LEFT, TURN\_RIGHT, RANDOM*

- Fitness function

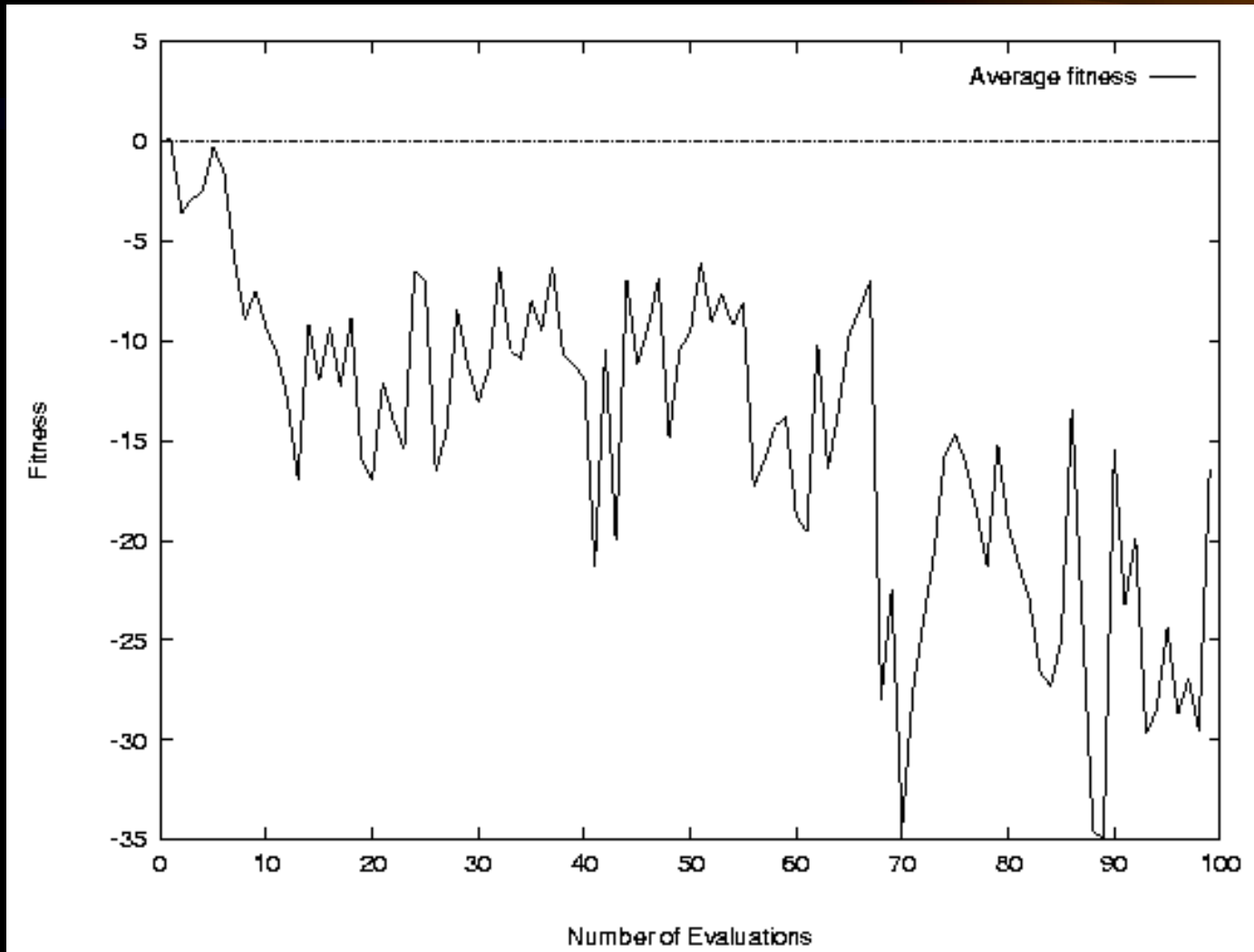
$$(1) F_{new} = F_{old} + w_1 \times (\#collisions) + w_2 \times (\#steps)$$

$$(2) F_{new} = F_{old} + w_1 \times (\#miss) + w_2 \times (\#steps) + vision \times w_3$$

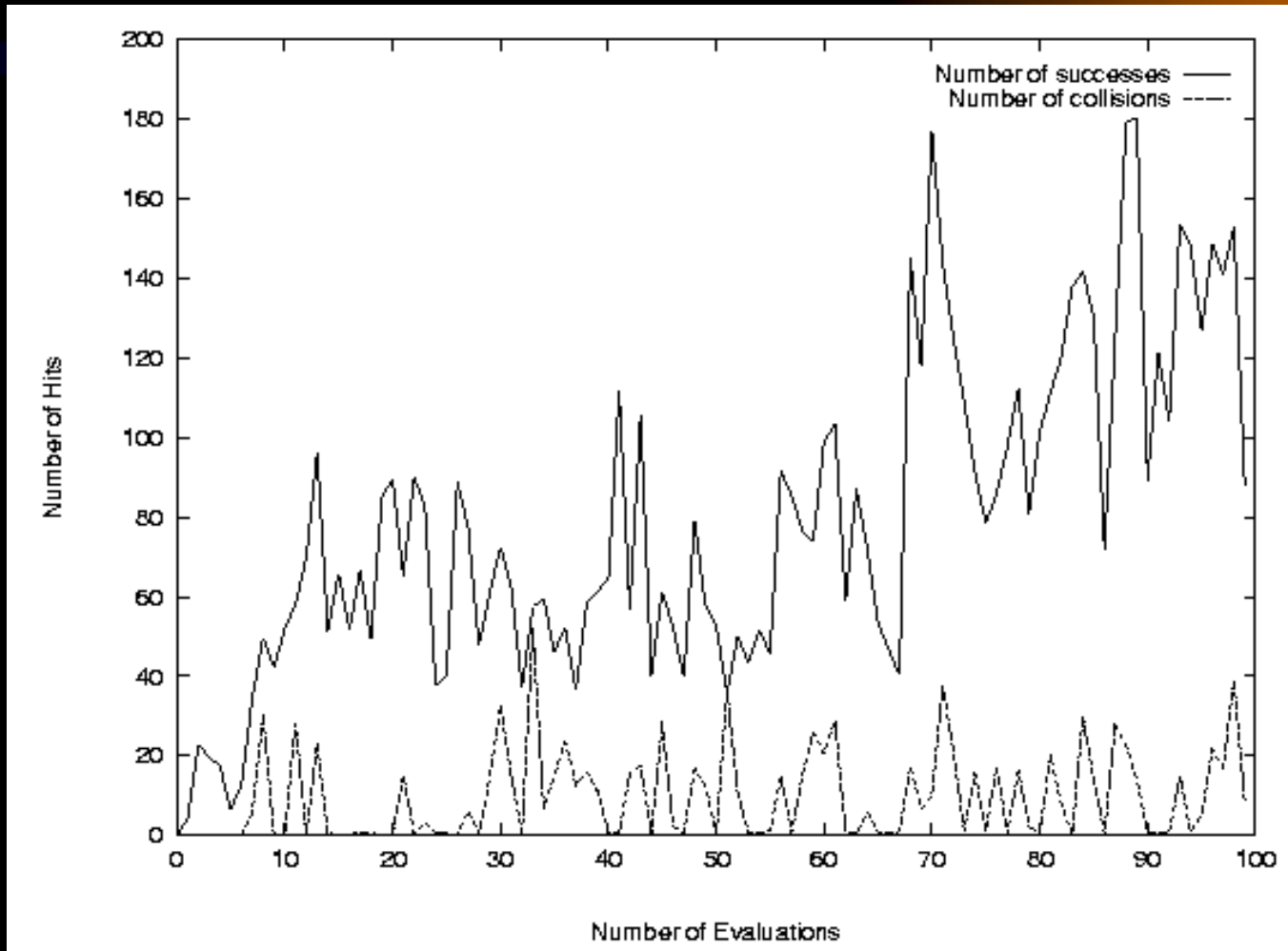
# *Experimental Results: Best Fitness*



# *Experimental Results: Average Fitness*



# *Experimental Results: Hits*





# *Conclusion and Future Work*

- Presented a method for evolving GP trees on EH
- Speed-up by reducing fitness evaluation time
- The larger the training set, the higher the speed-up factor
- Possibility of special-purpose GP hardware
- Appropriate for on-line hardware learning
- Further possibility of resource utilization in evolving GP trees on EH

# *Sources*



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