

Nature-Inspired Computing

Evolutionary Programming

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Overview

- developed in the USA in the 1960's
- typically applied to:
 - traditional EP: machine learning tasks by finite state machines
 - contemporary EP: (numerical) optimization
- features:
 - very open framework: any representation and mutation op's OK
 - crossbred with ES (contemporary EP)
 - consequently: hard to say what "standard" EP is
 - no recombination
 - self-adaptation of parameters standard (contemporary EP)

Overview (current EP)

Representation	Real-valued vectors
Recombination	None
Mutation	Gaussian mutation
Parent selection	Deterministic
Survivor selection	Probabilistic ($\mu+\mu$)
Specialty	Self-adaptation of mutation step sizes (in meta-EP)

Prediction by Finite State Machines

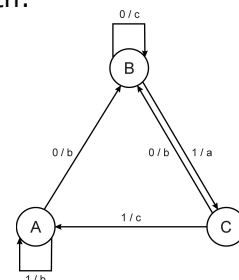
- finite state machine (FSM):
 - states S
 - inputs I
 - outputs O
 - transition function $\delta : S \times I \rightarrow S \times O$
 - transforms input stream into output stream

Prediction by Finite State Machines

- can be used for predictions, e.g. to predict next input symbol in a sequence

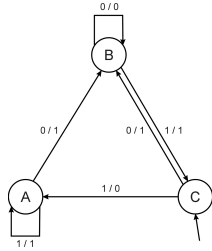
FSM example

- consider the FSM with:
 - $S = \{A, B, C\}$
 - $I = \{0, 1\}$
 - $O = \{a, b, c\}$
 - δ given by a diagram



FSM as Predictor

- consider the following FSM
- task: predict next input
- quality:
% of $in_{(i+1)} = out_i$
- given initial state C
- input sequence 011101
- leads to output 110111
- quality: 3 out of 5



Modern EP

- no predefined representation
- no predefined mutation (must match representation)
- often applies self-adaptation of mutation parameters