

Evolution Strategies Evolutionary Programming Genetic Programming

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Lecture Outline

- Evolutionary Algorithms revision
- Genetic Algorithms revision
- Evolution Strategies
- Evolutionary Programming
- Genetic Programming

Evolutionary Algorithms

- General purpose algorithms
- Based on some aspects of biological evolution
- Useful for solving multi-parameter optimisation problems
- NOT random search algorithms

Genetic Algorithms

- AKA GA
- Based on populations of artificial chromosomes
- Solution attempts are encoded as value strings in the chromosomes

Genetic Algorithms

- New individuals created by two mechanisms
 - crossover (recombination)
 - mutation
- Selection based on fitness of chromosomes
 - stochastic selection

Evolution Strategies

- Invented early 1960s in Germany
- Ingo Rechenberg, Hans-Paul Schwefel and Peter Bienert
 - engineering students
- Experimenting with wind tunnels
 - optimising jointed flat plates
- Only intuitive methods to do this at the time

Evolution Strategies

- Rechenberg had the idea of ‘mutating’ the parameters and selecting good mutations
- ES are used for numerical parameter optimisation
- Parameters of a problem are encoded as real numbers
 - real-numbered chromosome

Evolution Strategies

- Fitness of individual is determined by how well the parameters solve the problem
- Offspring are created by mutation
- Real-numbered, normally distributed creep mutation
- Offspring replace parents only if more fit

Evolution Strategies

- ES named according to number of parents and children at each generation
- 1+1 ES has one parent and one child

1+1 ES

1. Evaluate fitness of parent P , $f(P)$
2. Create child C by adding small normally distributed values to each parameter of P
3. Evaluate the fitness of C , $f(C)$
4. If $f(C) > f(P)$ then
replace P with C
5. Repeat Steps 2-4 until stopping condition

Evolution Strategies

- Later ES have populations
- (m+1) and (m,l) ES
- m (μ) is the size of the parent population
- l (λ) is the size of the offspring population
- offspring are created using recombination as well as mutation

Evolution Strategies

- In a (m+1) ES, the m best survive to the next generation
- In a (m,l) ES, only child individuals survive to the next generation

Evolutionary Programming

- Invented early 1960s in the USA
- Created by Lawrence Fogel
- Regarded artificial intelligence as the ability to predict a symbol based on previous symbols
- Evolved a population of finite state automata to perform this prediction

Evolutionary Programming

1. Create a population of solutions
2. Evaluate each solution in the population
3. Select individuals to reproduce
 - tournament selection
4. Mutate the reproduction population
5. Repeat 2 - 4 until stopping condition is reached

EP and GA

- Main differences between EP and GA
- GA requires the solution attempt to be encoded in a string of values
 - genome
- EP uses whatever representation fits the problem

Evolutionary Programming

- Evolution consisted of adding, modifying or deleting state transitions
- Task was to predict characters from streams of characters
- FSA with the least number of errors were allowed to reproduce

Evolutionary Programming

- Reproduction in EP is via mutation
- Mutation may be normally distributed
- No prescribed method of representation
 - Use whatever works for the problem
 - FSA, ANN etc
- Crossover / recombination is not used

EP and GA

- Mutation in EP is a normally distributed perturbation
- Has infrequent large changes, frequent small changes
- Mutation rate decays as run time elapses
- GA mutation tends to be fixed size changes that create entirely new values

EP and ES

- EP and ES are very similar
- Major differences are selection and recombination
- Selection in EP is stochastic
 - tournament based
 - randomly selected participants

EP and ES

- Selection in ES is deterministic
 - bad individuals are purged
 - good individuals breed
 - no randomness involved
- No recombination is used in EP
- Multi individual ES will use recombination

Genetic Programming

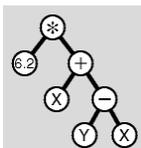
- Created by John Koza
- Use an EA to create algorithms
- Score attempts based on
 - how well the problem is solved
 - how efficiently it solves it

Genetic Programming

- Mechanistically similar to GA
- One major difference
 - no genotype / phenotype distinction
 - evolutionary operations carried out directly on the candidate programs themselves
- Reproduction is via crossover and mutation

Genetic Programming

- Candidates are represented as parse trees
- e.g.



Adapted from <http://www.genetic-programming.com>

Genetic Programming

- Crossover is implemented by swapping branches of the tree
- Mutation involves deleting and re-growing branches

Summary

- ES, EP and GP are all different kinds of evolutionary computation
- Each developed separately, but have common themes
- The boundaries between each are not clear-cut!
- Each have their own niches

References

- “Evolutionary Computation: Towards a New Philosophy of Machine Intelligence” by David B. Fogel
- “Genetic Programming” by John Koza