Genetic Algorithms
The Traditional Approach

- Ask an expert
- Adapt existing designs
- Trial and error
Nature’s Starting Point

Alison Everitt’s “A User’s Guide to Men”
Optimised Man!
Example: Pursuit and Evasion

- Using NNs and Genetic algorithm
- 0 learning
- 200 tries
- 999 tries
Comparisons

• Traditional
  • best guess
    • may lead to local, not global optimum

• Nature
  • population of guesses
    • more likely to find a better solution
More Comparisons

• Nature
  • not very efficient
    • at least a 20 year wait between generations
    • not all mating combinations possible

• Genetic algorithm
  • efficient and fast
    • optimization complete in a matter of minutes
    • mating combinations governed only by “fitness”
The Genetic Algorithm Approach

- Define limits of variable parameters
- Generate a random population of designs
- Assess “fitness” of designs
- Mate selection
- Crossover
- Mutation
- Reassess fitness of new population
A “Population”
Ranking by Fitness:
Mate Selection:
Fittest are copied and replaced less-fit
Mate Selection Roulette:
Increasing the likelihood but not guaranteeing the fittest reproduction
Crossover:
Exchanging information through some part of information (representation)
Mutation:
Random change of binary digits from 0 to 1 and vice versa (to avoid local minima)
Best Design
The GA Cycle
Genetic Algorithms

Adv:

• Good to find a region of solution including the optimal solution. But slow in giving the optimal solution
Genetic Approach

- When applied to strings of genes, the approaches are classified as genetic algorithms (GA)
- When applied to pieces of executable programs, the approaches are classified as genetic programming (GP)
- GP operates at a higher level of abstraction than GA
Typical “Chromosome”