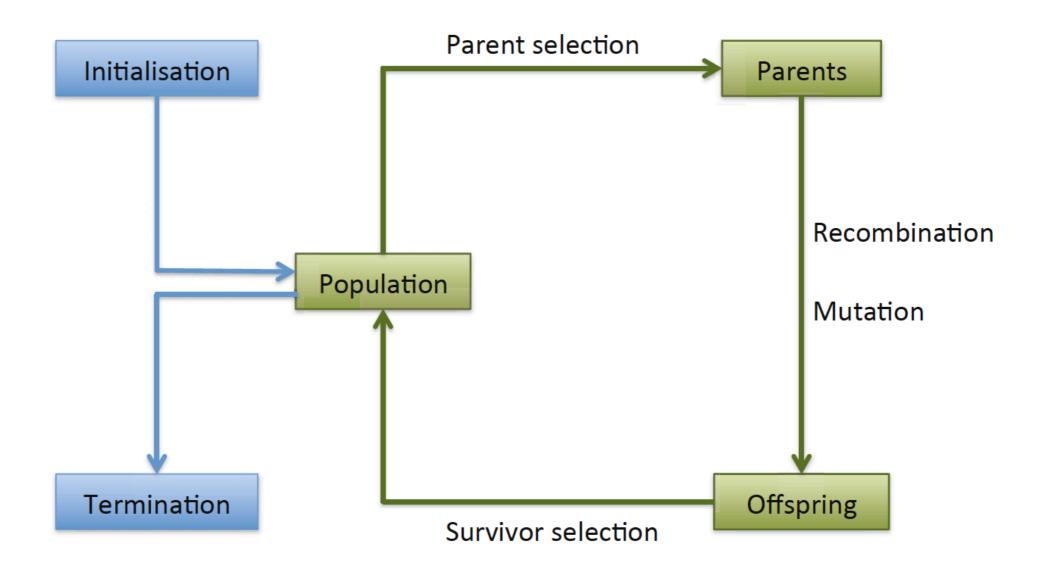
3: Evolutionary Algorithm Basics

- EA flowchart
- Representation
- Evaluation
- Population and selections
- Variation operators
- Textbook Chapter 3.2

General scheme of EAs



Main components of the design of an EA

- Representation (definition of individuals)
- Evaluation function (or fitness function)
- Population
- Parent selection mechanism
- Variation operators, recombination and mutation
- Survivor selection mechanism (replacement)

Pseudo-code for typical EA

```
BEGIN

INITIALISE population with random candidate solutions;

EVALUATE each candidate;

REPEAT UNTIL ( TERMINATION CONDITION is satisfied ) DO

1 SELECT parents;

2 RECOMBINE pairs of parents;

3 MUTATE the resulting offspring;

4 EVALUATE new candidates;

5 SELECT individuals for the next generation;

OD

END
```

Representations

- Candidate solutions (individuals) exist in phenotype space
- They are encoded in chromosomes, which exist in genotype space
 - Encoding: phenotype -> genotype
 - Decoding: genotype -> phenotype (one to one)
- Chromosomes contain genes, which are in (usually fixed) positions called loci and have a value (allele)

In order to find the global optimum, every feasible solution must be represented in genotype space

Evaluation (fitness) function

- Represents the requirements that the population should adapt to
 - a.k.a. quality function or objective function
- Assigns a single real-valued fitness to each phenotype which forms the basis for selection
- Typically we talk about fitness being maximized
 - Some problems may be best posed as minimization problems

Implicit fitness

- For many of the approaches, an explicit fitness function is used. Why?
 - better determine the kinds of good solutions
 - useful to solve particular problems
- Can have fitness be determined implicitly
 - competition in some form for resources, where better competitors are more fit
 - Tierre, Avida, mostly Artificial Life

Population

- Holds (representations of) possible solutions
- Usually has a fixed size and is a multiset of genotypes
- Some sophisticated EAs also assert a spatial structure on the population, i.e. a grid
- Selection operators usually take whole population into account, i.e reproductive probabilities are relative to *current* generation
- Diversity of a population refers to the number of different fitnesses / phenotypes / genotypes present (note not the same thing)

Parent selection mechanism

- Assigns variable probabilities of individuals acting as parents depending on their fitnesses
- Usually probabilistic
 - High quality solutions more likely to become parents than low quality ones
 - but not guaranteed
 - even worst in current population usually has non-zero probability of becoming a parent
- This stochastic nature can aid escape from local optima but slow you down

Variation operators

- Role is to generate new candidate solutions
 - From parents to child (offspring)
- Usually divided into two types according to their arity (number of inputs):
 - Arity I: mutation operator
 - Arity > I: recombination operator
 - Arity = 2: typically called crossover
- There has been much debate about relative importance of recombination and mutation
 - Nowadays most EAs use both
 - Choice of particular variation operators is representation dependent

Mutation

- Act on one genotype and delivers another
- Element of randomness is essential and differentiates it from other unary heuristic operators
- Importance ascribed depends on representation and dialect:
 - Binary GAs background operator responsible for preserving and introducing diversity
 - Continuous variables EP sole search operator
- May guarantee connectedness of search space and hence coverage proofs

Recombination

- Merges information from parents into offspring
- Choice of what information to merge is stochastic
- Most offspring may be worse, or the same as the parents
 - Hope is that some are better by combining elements of genotypes that lead to good traits
- In nature, higher-level organisms reproduce sexually while lower-level ones reproduce asexually recombination is the superior form of reproduction

Survivor selection (replacement)

- Sometimes also called environmental selection
- Most EAs use fixed population size, and need a way of going from (parents + offspring) to next generation
- Often deterministic (different from parent selection!)
 - Fitness based: e.g., ranking parents + offspring and save the best
 - Age based: e.g., use offspring to replace worst individuals in the previous generation
- Elitism

Initialization and termination

- Initialization usually done at **random**
 - Need to ensure even spread and mixture of possible allele values
 - Can include existing solutions, or use problem-specific heuristics to "seed" the population
- Termination condition checked every generation
 - Reaching some (known/hoped for) fitness
 - Reaching some allowed limit number of generations
 - Reaching some minimum level of diversity
 - Reaching some specified number of generations without fitness improvement (stagnation)
 - Anytime algorithm

General scheme of EAs

