



On Differential Evolution

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Biologically-Inspired Intelligent Systems

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Evolutionary Computation through GAs

Genetic Algorithm

Objective function $f(x)$, $x = (x_1, \dots, x_d)^T$

Encode the solutions into chromosomes (strings)

Define fitness F (eg, $F \propto f(x)$ for maximization)

Generate the initial population

Initialize the probabilities of crossover (p_c) and mutation (p_m)

while ($t <$ Max number of generations)

 Generate new solution by crossover and mutation

 Crossover with a crossover probability p_c

 Mutate with a mutation probability p_m

 Accept the new solutions if their fitness increase

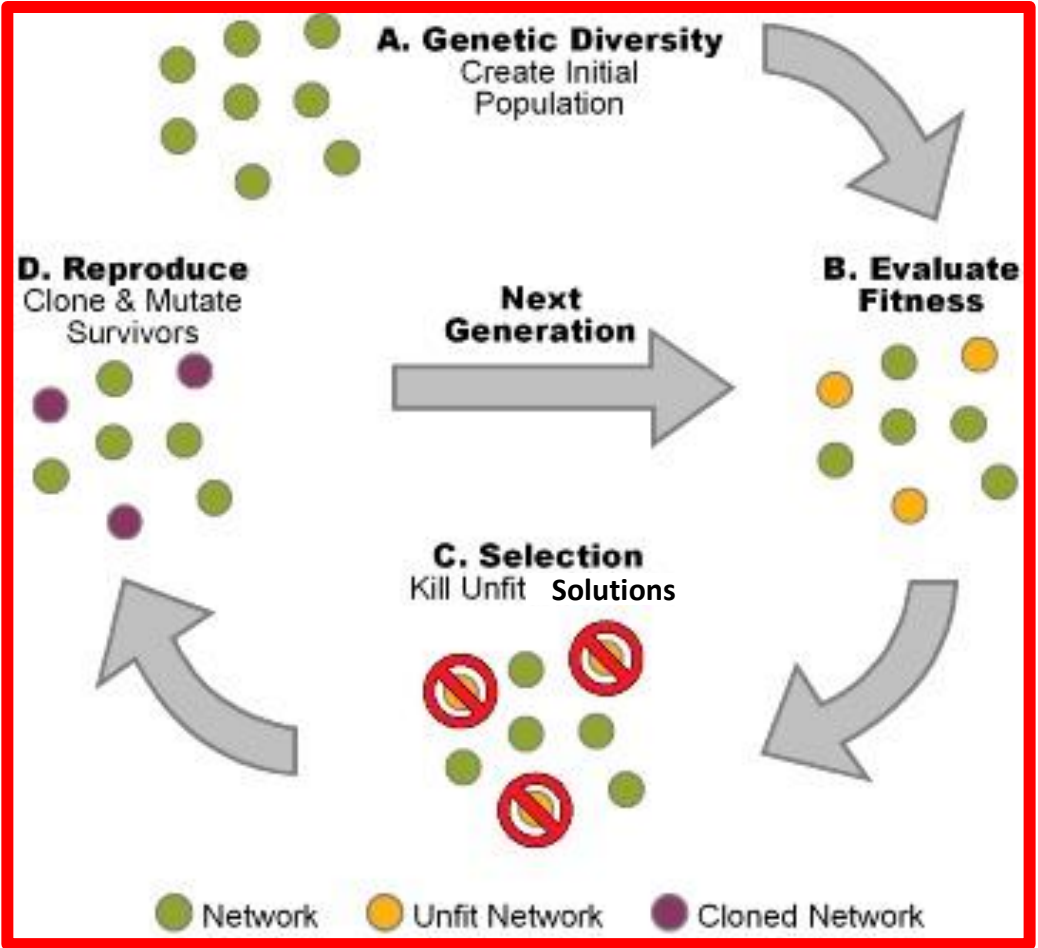
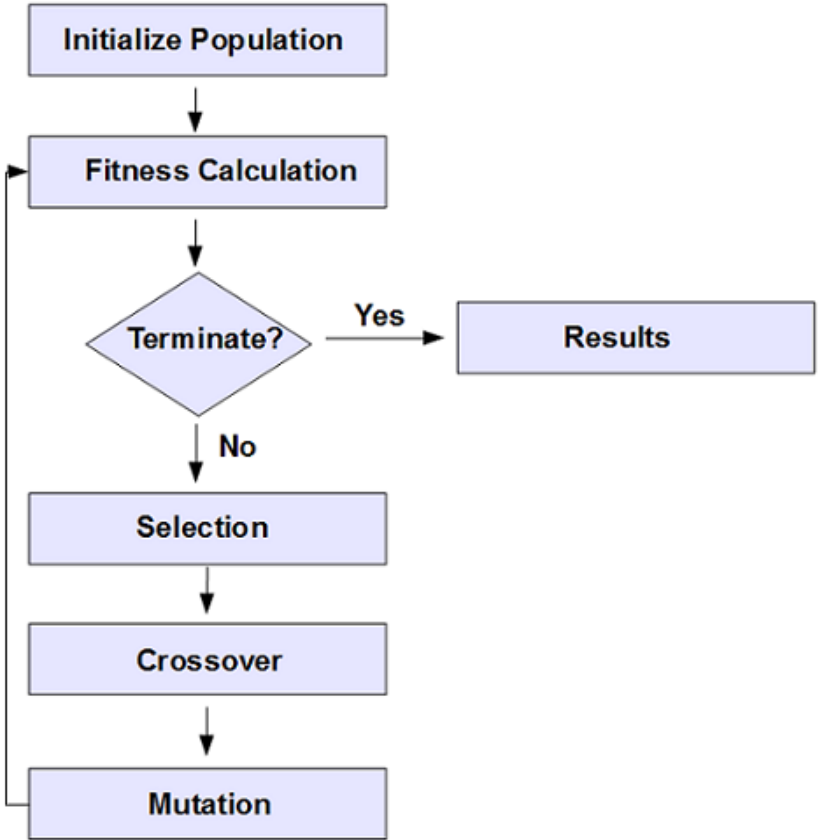
 Select the current best for the next generation (elitism)

 Update $t = t + 1$

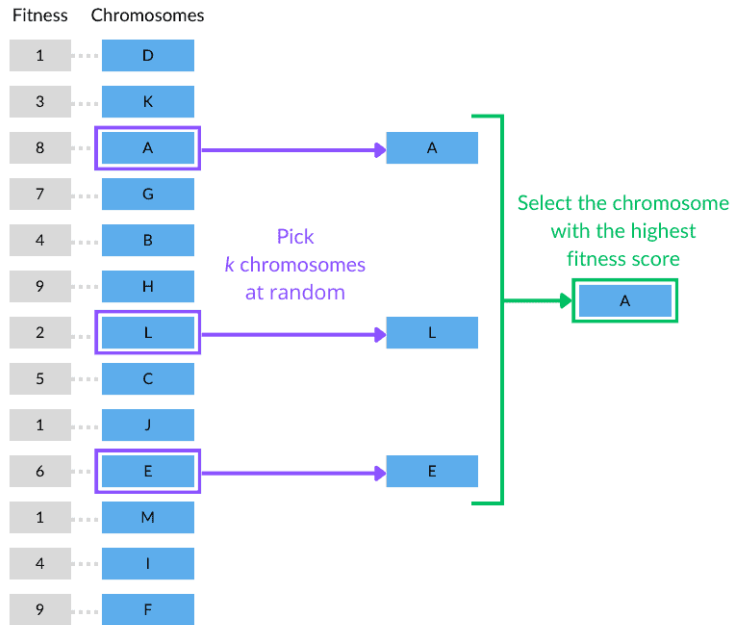
end while

Decode the results and visualization

Genetic Algorithm (Process Flow)



Tournament Selection



```
/* Select the fittest individual (or chromosome) from a randomly selected list of individuals T */
algorithm Tournament(T, k):
  // INPUT
  // T = a list of individuals randomly selected from a population.
  // k = the tournament size. In other words, the number of elements in T.
  // OUTPUT
  // the fittest individual.
  Best ← T[1];
  for i from 2 to k do
    Next ← T[i];
    if Fitness(Next) > Fitness(Best) then
      Best ← Next;
  return Best;
```

```
/* Assume we wish to select n individuals from the population P */
algorithm TournamentSelection(P, k, n):
  // INPUT
  // P = the population.
  // k = the tournament size, such that  $1 \leq k \leq$  the number of individuals in P.
  // n = the total number of individuals we wish to select.
  // OUTPUT
  // the pool of individuals selected in the tournaments.
  T ← an empty sequence;
  B ← an empty sequence;
  for i from 1 to n do
    Pick k individuals from P at random, with or without replacement, and add them to T;
    B[i] ← Tournament(T, k);
    T ← [ ]
  return B;
```

Procedure Memetic Algorithm Based on an EA

Initialization: $t = 0$; // Initialization of the generation counter

Randomly generate an initial population $P(t)$;

Compute the fitness $f(p) \forall p \in P(t)$;

while Stopping conditions are not satisfied **do**

Selection: Accordingly to $f(p)$ choose a subset of $P(t)$ and store it in $M(t)$;

Offspring: Recombine and mutate individuals $p \in M(t)$ and store them in $M'(t)$;

Learning: Improve p' by local search or heuristic $\forall p' \in M'(t)$;

Evaluation: Compute the fitness $f(p') \forall p' \in M'(t)$;

if Lamarckian learning **then**

Update chromosome of p' according to improvement $\forall p' \in M'(t)$;

fi

New generation: Generate $P(t+1)$ by selecting some individuals from $P(t)$ and $M'(t)$;

$t = t + 1$; // Increment the generation counter

end while

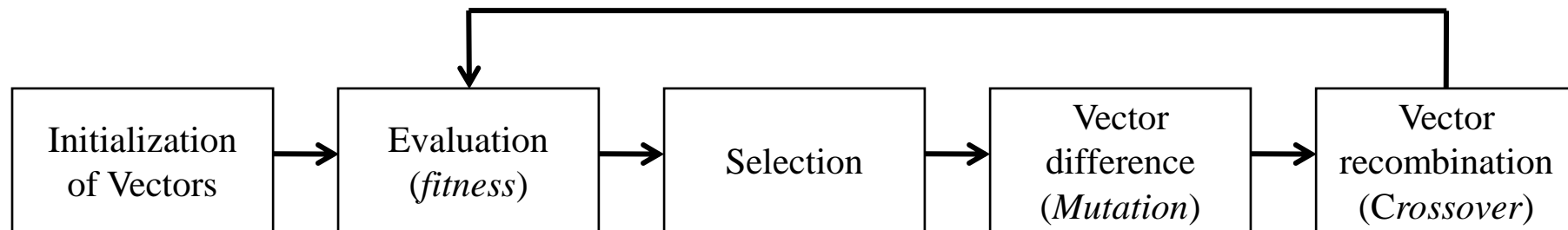
Return best individual $p \in P(t-1)$ as result;

Genetic Algorithms: Race Car Evolution

- <https://www.youtube.com/watch?v=Aut32pR5PQA>

Metaheuristic: Differential Evolution (DE)

- Vector-based (population-based) algorithm, Storn & Price (1996/1997)
 - Individuals 'evolve' by recombination w/ other individuals & differentials between other individuals
- Devised for continuous search spaces, derivative-free
- No encoding/decoding required – real numbers are solution strings/chromosomes
- *DE/rand/1/bin*



DE General Mechanics

- Builds on the idea of genetic algorithms
- Three primary steps:
 - Mutation, crossover, selection
- Name convention: DE/x/y/z – x is mutation scheme, e.g., random (Rand) or best (Best), y is number of difference vectors, z is crossover scheme, e.g., binomial (Bin) or exponential (Exp) or either/agnostic (*)
 - We will start with: DE/Rand/1/*

Questions?

