
MDL Assignment-3

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1 File Structure

Upon running the file *solution.py*, we get

```
team71
├── report.pdf:  contains report
├── client.py:   contains necessary functions for server
├── set_cur_state.py: contains necessary functions for vectors
├── solution.py: main solution
└── trace_71.txt: trace obtained
```

2 Problem Statement

Given coefficients of vectors corresponding to over-fit model, apply Genetic Algorithm (G.A) to reduce over-fitting.

Input:

$\bar{v}_{overfit} = [-0.00016927573251173823, 0.0010953590656607808,$
 $0.003731869524518327, 0.08922889556431182,$
 $0.03587507175384199, -0.0015634754169704097,$
 $-7.439827367266828e-05, 3.7168210026033343e-06,$
 $1.555252501348866e-08, -2.2215895929103804e-09,$
 $2.306783174308054e-11]$

Given a query server to allow for submission of weights (w_i), the following is returned for each query -

$$M.S.E = \frac{\sum_{x \in data} (y - f(x))^2}{N}$$

where $f(x)$ is calculated using the weights submitted $[w_1, w_2 \dots w_{10}]$ -

$$f(x) = \sum_{i=1}^{10} w_i \cdot x_i$$

Output: A set of weights $[w_1, w_2 \dots w_{10}]$ corresponding to best-fit under the conditions-

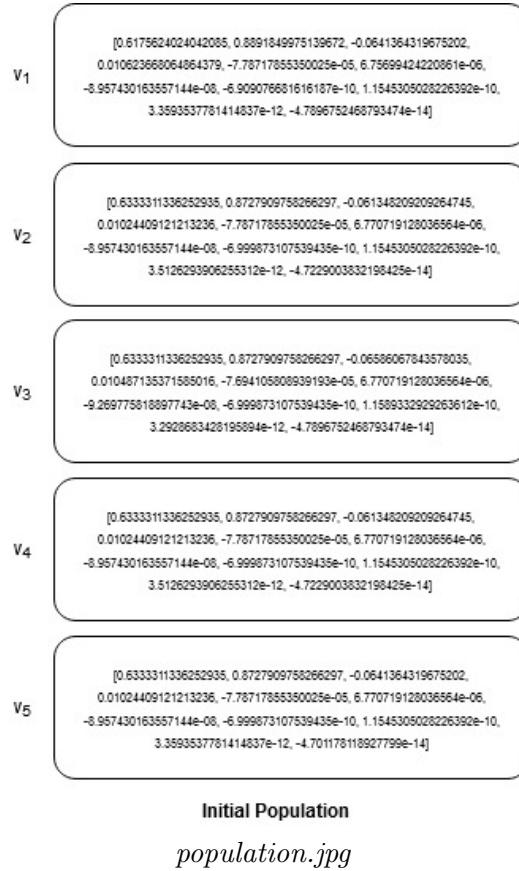
$$|w_i| \leq 10$$

3 Algorithm Definition

Our Genetic Algorithm consists of 4 steps -

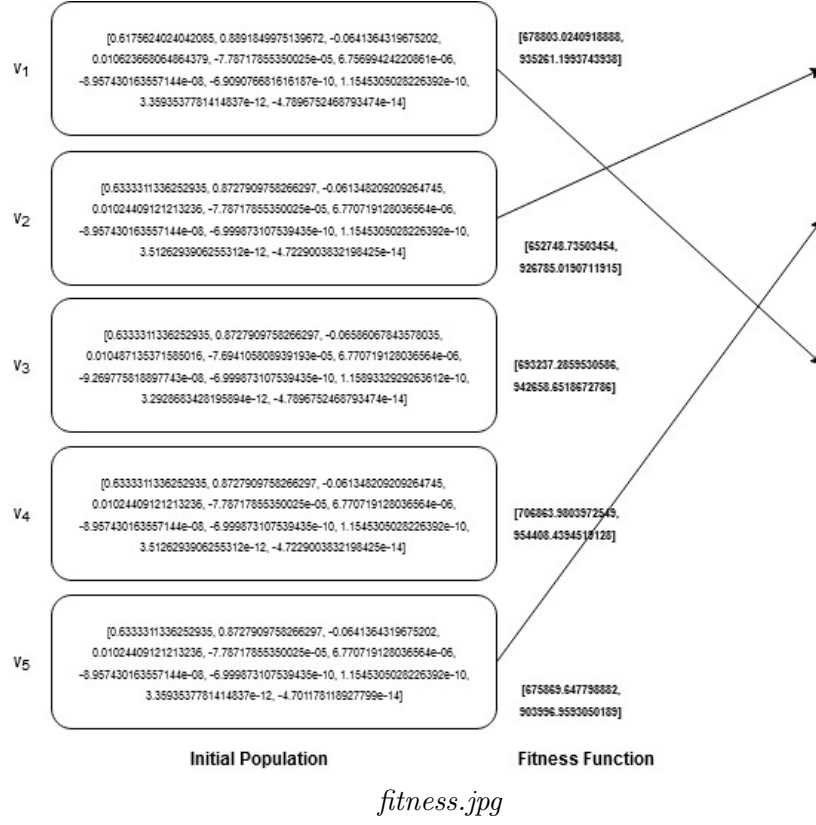
1. **Initialisation of Population:**

Each generation contains 5 vectors. This is chosen from previous set of population or derived initially randomly. After the algorithm runs, this population is instead chosen from the already saved state in *current.state.pkl*.



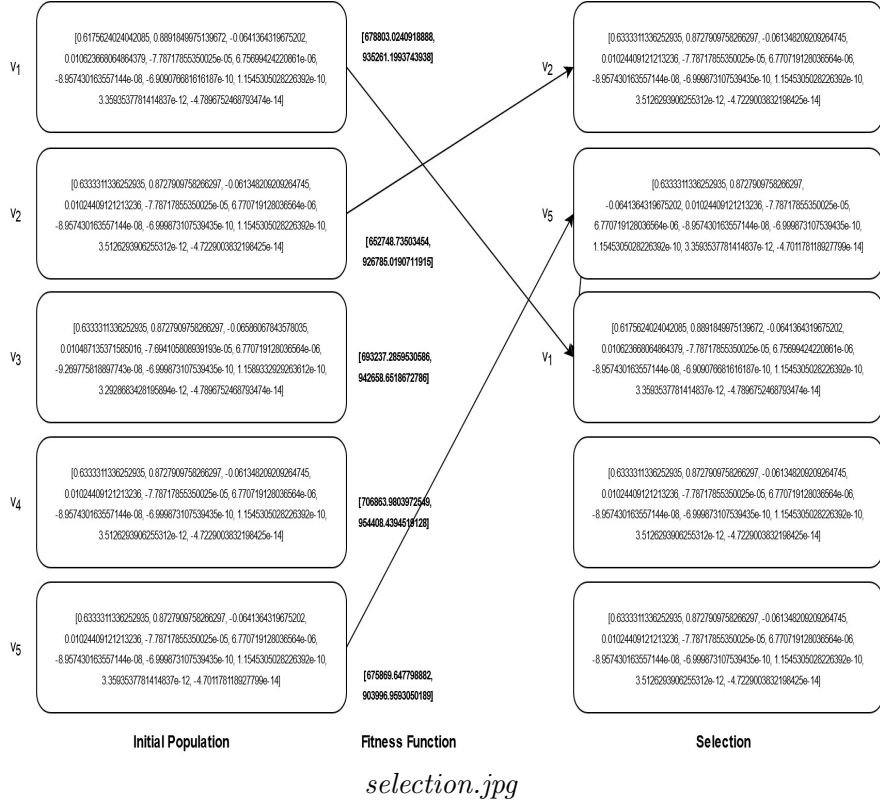
2. **Fitness Function:**

The best three of each iteration survive and give birth to next generation. We chose training error to be out fitness function.



3. Selection

Since we have used training error as our fitness function , the best 3 vectors are **deterministically** chosen from the population. This is where we differ from the normal Genetic Algorithm that the selection is not random.



So if A, B, C, D, E are in sorted order of their fitness function then A, B, C survive. The best three of each iteration survive and give birth to next generation.

4. Crossover with Mutation

Crossover

We did a probabilistic cross over which is explained below.

A is best vector and B is second best vector, then let Crossover vector be C .

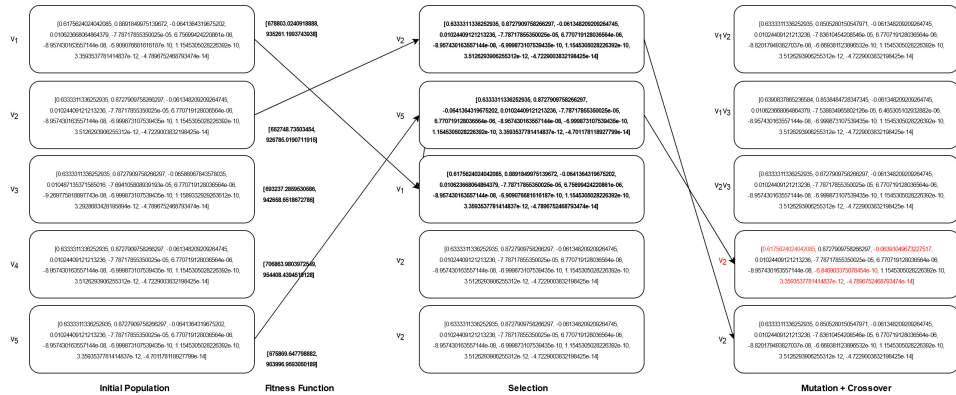
$$C_i = \begin{cases} A_i, & \text{with } p_i = 0.6 \\ B_i, & \text{with } q_i = 0.4 \end{cases}$$

Mutation The mutation was also probabilistic. Let M be the new mutation vector.

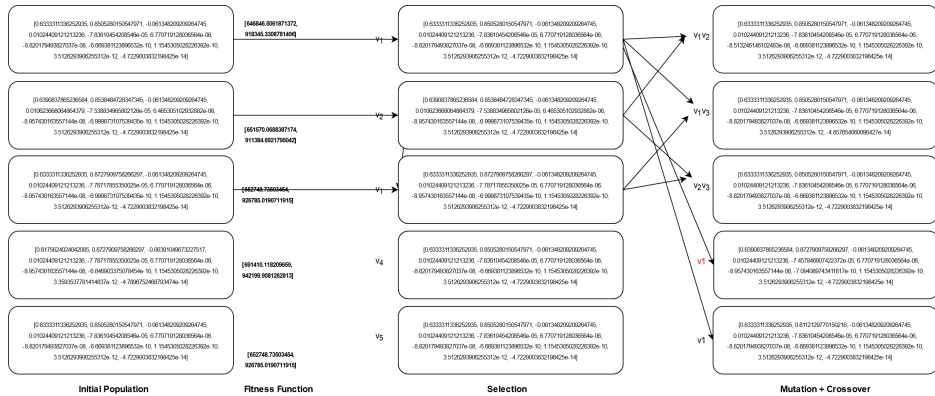
$$M_i = \begin{cases} A_i, & \text{with } p_i = 0.6 \\ \in (0.95 * A_i, 1.05 * A_i), & \text{with } q_i = 0.4 \end{cases}$$

Suppose the chosen population was A,B and C (with A giving the best value in fitness function, B the next best and then C). The new vectors are

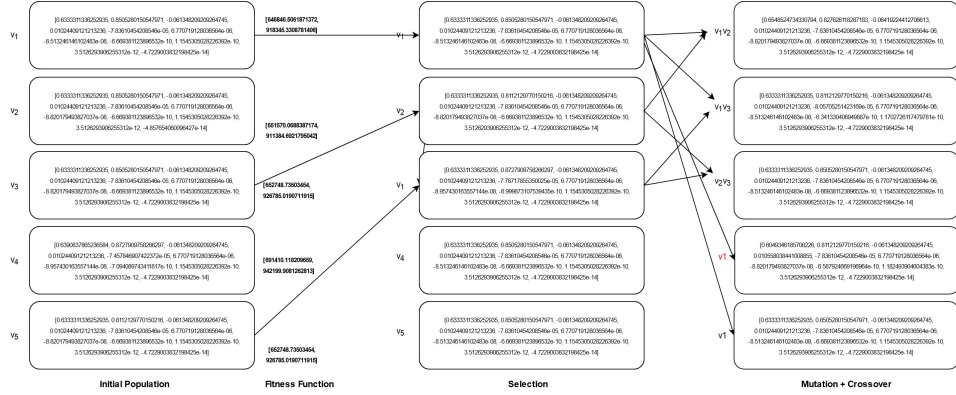
- A-B (Crossover with Mutation).
- A-C (Crossover with Mutation).
- B-C (Crossover with Mutation).
- A with mutations: This is A with the following mutation.
- A without any mutation: This is just plain A without any mutation.



iteration1.jpg



iteration2.jpg



iteration3.jpg

The heuristics We did a probabilistic cross over which is explained above. We introduced the concept of survivability The best vector always survive. The best vector also produces a copy of itself with some mutations. This helps in converging towards a minimum. Also as the error becomes less we did a trick. (A-B)[0] survives only if it is better than vector of previous generation.