

A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light greenish-blue. They are positioned diagonally, with the blue one partially covering the green one.

# Evolutionary Computation Sort

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# Background

## Problem

- Sort a list of integers in ascending order.
- The search space for this problem are all permutations of the given list.
  - This gives us a search space of  $n!$ , where  $n$  is the size of the list to sort.

## Genetic Algorithm Framework

- Common steps
  - 1) generate a population
  - 2) generate offspring
  - 3) reduce the population to original size

```
1 #!/usr/bin/env python
2 import random
3 import itertools
4 from Individual import Individual
5
6 class GeneticAlgorithmSort(object):
7     def sort(self):
8         generation_count = self.generations
9         #create initial m population
10        self.createInitialPopulation()
11        #keep evolving while not max generations (iterations) or sorted list found
12        while (generation_count > 0) and self.population[0].fitness() < self.optimalFitness:
13            #update generation count
14            generation_count -= 1
15            # reset offspring list
16            self.offspring = []
17            # generate n offspring (parents should be same size as offspring size)
18            # select parent(s) (a list of individual or individual couples)
19            parents = self.parentSelect(self.population, self.offspringSize)
20            for parent_s in parents:
21                # produce offspring (crossover, etc.) reproduce returns a list
22                offspring = self.reproduce(parent_s)
23                offspring = self.mutation(offspring)
24                # mutate (random swap)
25                self.offspring.append(Individual(offspring, self.fitness))
26            #reduce population from m+n down to n
27            self.population = self.selection(self.population, self.offspring, self.populationSize)
28            # return sorted list if while has exited
29            return self.population[0].getIndividual()
30
31        def createInitialPopulation(self):
32            # add initial list to population
33            self.population = [Individual(self.original_list, self.fitness)]
34            # generate permutation of original list
35            permutations = list(itertools.permutations(self.original_list))
36            # randomly select m individuals from the permutation list to add to the population
37            for i in range(self.populationSize-1):
38                rand_idx = int( random.random()*len(permutations) )
39                self.population.append(Individual(permutations[rand_idx], self.fitness))
```

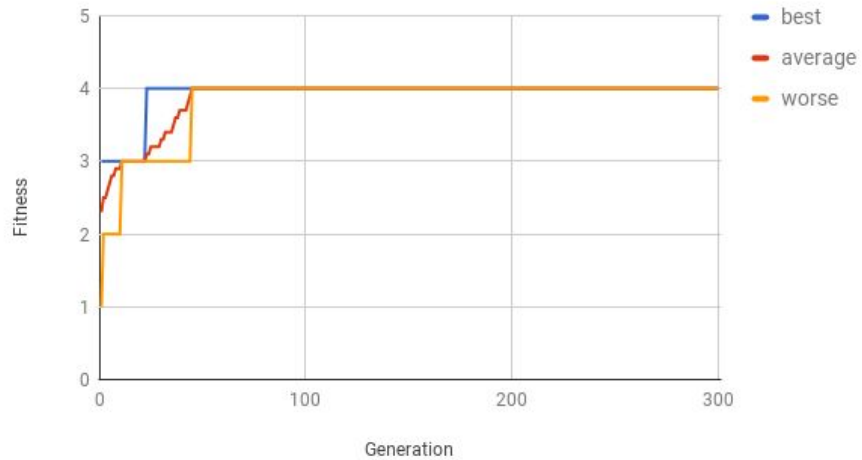


# Tunable Methods

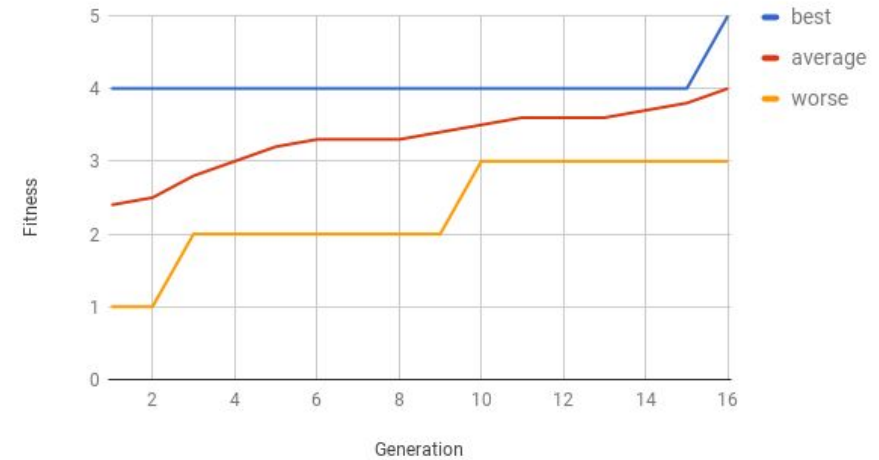
- Fitness Methods
  - Number of correctly sorted values based on the first value
  - E.g [3, 1, 2, 5, 4]
- Parent Selection Methods
  - Fittest individual from population
- Reproduction Methods
  - Clone parent
- Mutation Methods
  - Randomly swap 2 indices from the list
- Selection Methods
  - Truncation of fittest individuals

# Results

Fitness vs Generation (close solution found)



Fitness vs Generation for successful solve





# Conclusion

- Maybe genetic algorithms are not good for sorting problems
  - genetic algorithm guarantee a close solution but not the optimal solution
- Exploring different ways of producing offspring would be the next step
  - Parent selection
  - Reproduction
  - Mutation methods