## Project for Casimir Programming Course

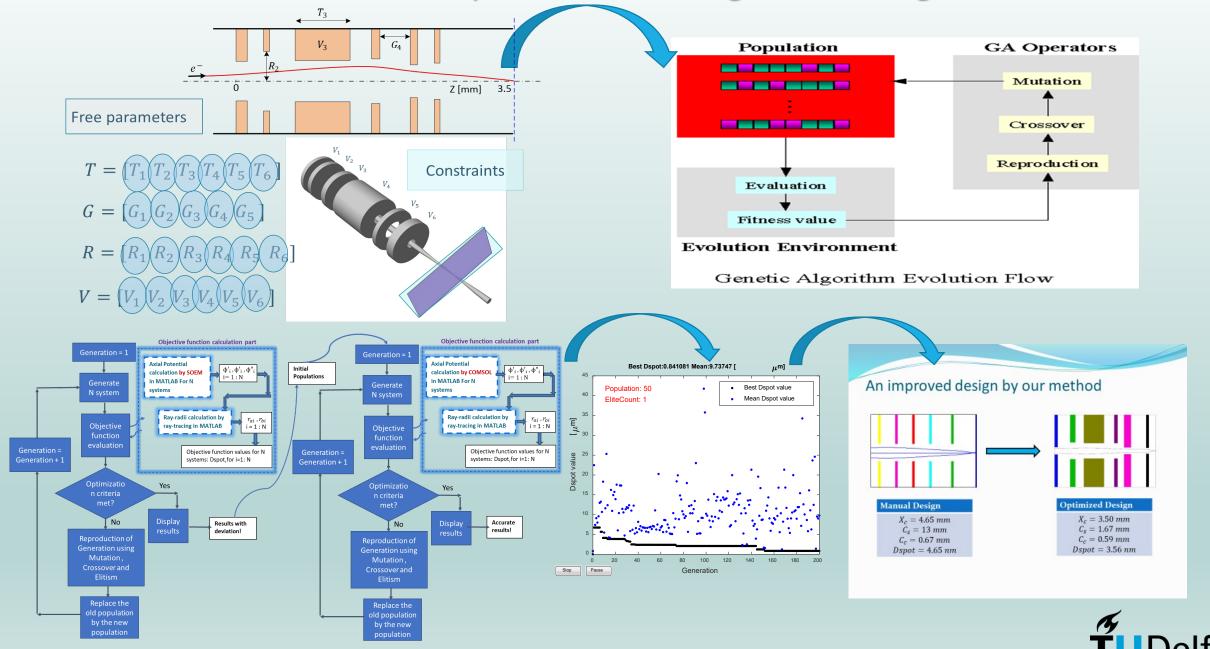
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8/11/2019

TUDelft- Applied Physics
Imaging Physics
Charged Particle Optics



### **Electrostatic Lens Optimization Using Genetic Algorithms**



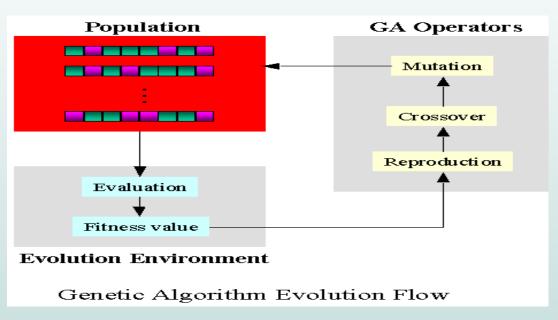
# Optimization using Genetic Algorithms By Python!?

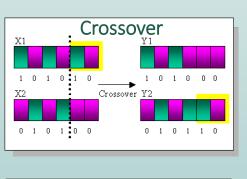
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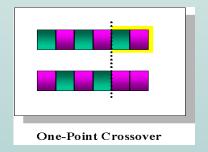
**Continuous Genetic Algorithm From Scratch With Python!** ©

https://towardsdatascience.com/continuous-genetic-algorithm-from-scratch-with-python-ff29deedd099

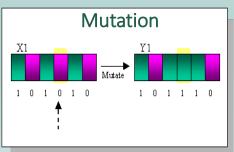






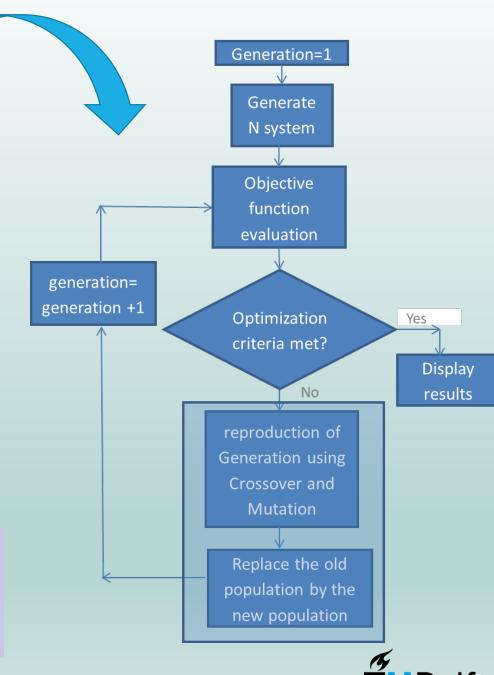














```
def individual(number of genes, upper limit, lower limit):
    individual=[round(rnd()*(upper limit-lower limit)
                 +lower limit,1) for x in range(number of genes)]
    \#round(a,n) round the values of a to n digits (here to one digit),
    \#rnd()=random\ number\ from\ [0,1)\ (i.e.\ =generates\ Random\ float\ x,\ 0.0\ <=\ )
    return individual
print(individual(4,1,10)) #the upper and lower limit should be integers not
[7.2, 6.1, 4.3, 8.4]
def(population)number_of_individuals,
                number_of_genes, upper_limit, lower_limit):
    return
        individual(number of genes, upper limit, lower limit)
        for x in range(number of individuals)
print(population(8,4,10,1))
def fitness calculation(individual):
    fitness value = sum(individual)
   return fitness_value
   # print(fitness value)
print(fitness calculation([5.5, 3.8, 3.4, 9.9]))
[4.0, 8.1, 7.5, 1.2], [8.4, 9.7, 8.1, 3.0], [8.1, 2.4, 4.6, 2.2], [9.2, 8.2, 8.9, 1.7],
[4.0, 6.7, 3.2, 5.7], [8.6, 3.1, 8.6, 2.5], [7.7, 4.3, 1.8, 6.8], [2.6, 5.7, 8.2, 1.5]]
22.6
def roulette(cum_sum, chance):
    veriable = list(cum sum.copy())
    veriable.append(chance)
    veriable = sorted(veriable)
    return veriable.index(chance)
```

import numpy as np

from numpy.random import random as rnd

from random import gauss, randrange

-Calling functions from python library, such as numpy, randint, gauss

- -Define function "individual" to create Individuals
- Input: "number of Genes", "upper limit", "lower limit"
- Output: an "Individual"

- -Define function "Population" to create many Individuals
- Input: "number of individuals", "number of Genes", "upper limit", "lower limit"
- Output: N "Individual"s
- -Define function "Fitness Calculation" to calculate fitness
- Input: "individual"
- Output: fitness value of the individual

	Generation				
	Gene 1	Gene 2	Gene 3	Gene 4	Fitness
Individual 1	10	1	9	9	29
Individual 2	8	8	5	7	28
Individual 3	10	5	6	6	27
Individual 4	8	5	7	4	24
Individual 5	4	4	6	3	17
Individual 6	2	5	5	4	16
Individual 7	8	0	6	0	14
Individual 8	4	3	2	1	10



```
def (selection ()generation, method='Fittest Half'):
    generation['Normalized Fitness'] = \
        sorted([generation['Fitness'][x]/sum(generation['Fitness'])...
        for x in range(len(generation['Fitness']))], reverse = True)
    # this three-lines is one large line . \ is used when we want to continue the sentence
    generation['Cumulative Sum'] = np.array(
        generation['Normalized Fitness']).cumsum()
    if method == 'Roulette Wheel':
        selected = []
        for x in range(len(generation['Individuals'])//2):
            selected.append(roulette(generation
                ['Cumulative Sum'], rnd()))
            while len(set(selected)) != len(selected):
                selected[x] = \
                    (roulette(generation['Cumulative Sum'], rnd()))
        selected = {'Individuals':
            [generation['Individuals'][int(selected[x])]
                for x in range(len(generation['Individuals'])//2)]
                ,'Fitness': [generation['Fitness'][int(selected[x])]
                for x in range(
                    len(generation['Individuals'])//2)]}
   elif method == 'Fittest Half':
        selected individuals = [generation['Individuals'][-x-1] #start to count from end of vector -1
            for x in range(int(len(generation['Individuals'])/(2))] #// means round and make it intiger
        selected_fitnesses = [generation['Fitness'][-x-1]
            for x in range(int(len(generation['Individuals'])//2))]
        selected = {'Individuals': selected_individuals,
                    'Fitness': selected fitnesses}
   elif method == 'Random':
        selected individuals = \
            [generation['Individuals']
                [randint(1,len(generation['Fitness']))]
            for x in range(int(len(generation['Individuals'])//2))]
        selected fitnesses = [generation['Fitness'][-x-1]
            for x in range(int(len(generation['Individuals'])//2))]
        selected = {'Individuals': selected individuals,
                    'Fitness': selected fitnesses}
   return selected
id1 = individual(4,1,10)
generation={"Fitness": fitness calculation(id1)}
print (selection(generation, 'Fittest Half'))
```

- -Define function "Selection" to select best Individuals to bring into next generation
- Input: "individuals", "generation", "method of selection"
- Output: N best selected "Individual"s



```
def(pairing)elit, selected, method = 'Fittest'):
    individuals = [elit['Individuals']]+selected['Individuals']
    fitness = [elit['Fitness']]+selected['Fitness']
    if method == 'Fittest':
        parents = [[individuals[x],individuals[x+1]]
                  for x in range(len(individuals)//2)]
    if method == 'Random':
        parents = []
        for x in range(len(individuals)//2):
            parents.append(
                [individuals[randint(0,(len(individuals)-1))],
                individuals[randint(0,(len(individuals)-1))]])
            while parents[x][0] == parents[x][1]:
                parents[x][1] = individuals[
                    randint(0,(len(individuals)-1))]
    if method == 'Weighted Random':
        normalized fitness = sorted(
            [fitness[x] /sum(fitness)
             for x in range(len(individuals)//2)], reverse = True)
        cummulitive sum = np.array(normalized fitness).cumsum()
        parents = []
        for x in range(len(individuals)//2):
            parents.append(
                [individuals[roulette(cummulitive sum,rnd())],
                individuals[roulette(cummulitive sum,rnd())]])
            while parents[x][0] == parents[x][1]:
                parents[x][1] = individuals[
                    roulette(cummulitive sum,rnd())]
    return parents
```

-Define function "pairing" to choose how many of the best Individuals we want to keep and bring into next generation

- Input: selected N "individuals", "elit", "method of pairing"
- Output: N new "Individual"s ,called "parents"



```
def mating(parents, method='Single Point'):
    if method == 'Single Point':
        pivot point = randint(1, len(parents[0]))
       offsprings = [parents[0] \
            [0:pivot point]+parents[1][pivot point:]]
       offsprings.append(parents[1]
           [0:pivot point]+parents[0][pivot point:])
    if method == 'Two Pionts':
       pivot point 1 = randint(1, len(parents[0]-1))
       pivot point 2 = randint(1, len(parents[0]))
       while pivot point 2<pivot point 1:
            pivot point 2 = randint(1, len(parents[0]))
       offsprings = [parents[0][0:pivot_point_1];
           parents[1][pivot point 1:pivot point 2]+
            [parents[0][pivot point 2:]]]
       offsprings.append([parents[1][0:pivot point 1]+
           parents[0][pivot point 1:pivot point 2]+
            [parents[1][pivot_point_2:]]])
   return offsprings
def mutation()individual, upper limit, lower limit, muatation rate=2,...
    method= Reset', standard deviation = 0.001):
    gene = [randint(0, 7)]
    for x in range(muatation rate-1):
         gene.append(randint(0, 7))
        while len(set(gene)) < len(gene):</pre>
             gene[x] = randint(0, 7)
    mutated_individual = individual.copy()
    if method == 'Gauss':
        for x in range(muatation rate):
             mutated individual[x] = \
             round(individual[x]+gauss(0, standard deviation), 1)
    if method == 'Reset':
        for x in range(muatation rate):
             mutated_individual[x] = round(rnd()*_\
                 (upper limit-lower limit)+lower limit,1)
     return mutated individual
```

- -Define function "mating" to combine (crossover), the individuals to create new individuals
- Input: N selected "individuals"
   (=parents), "method of mating"
- Output: N new "individulas" (children)
  , called "offsprings"

- -Define function "mutation" to insert some mutated individuals among others
- Input: N selected "individuals", "method of mutation"
- Output: N "mutated individuals"



```
def next_generation(gen, upper_limit, lower_limit):
    elit = {}
    next gen = \{\}
    elit['Individuals'] = gen['Individuals'].pop(-1)
    elit['Fitness'] = gen['Fitness'].pop(-1)
    selected = selection(gen)
    parents = pairing(elit, selected)
    offsprings = [[[mating(parents[x])
                   for x in range(len(parents))]
                   [y][z] for z in range(2)]
                   for y in range(len(parents))]
    offsprings1 = [offsprings[x][0]
                  for x in range(len(parents))]
    offsprings2 = [offsprings[x][1]
                  for x in range(len(parents))]
   unmutated = selected['Individuals']+offsprings1+offsprings2
   mutated = [mutation(unmutated[x], upper_limit, lower_limit)...
       for x in range(len(gen['Individuals']))]
   unsorted individuals = mutated + [elit['Individuals']]
   unsorted next gen = \
        [fitness calculation(mutated[x])
        for x in range(len(mutated))]
   unsorted fitness = [unsorted next gen[x]
        for x in range(len(gen['Fitness']))] + [elit['Fitness']]
    sorted next gen = \
         for x in range(len(mutated))]
    unsorted fitness = [unsorted next gen[x]
        for x in range(len(gen['Fitness']))] + [elit['Fitness']]
    sorted next gen = \
        sorted([[unsorted_individuals[x], unsorted_fitness[x]]
            for x in range(len(unsorted individuals))],
                key=lambda x: x[1])
    next gen['Individuals'] = [sorted next gen[x][0]
        for x in range(len(sorted_next_gen))]
    next gen['Fitness'] = [sorted next gen[x][1]
        for x in range(len(sorted next gen))]
    gen['Individuals'].append(elit['Individuals'])
    gen['Fitness'].append(elit['Fitness'])
    return next gen
```

#### -Define function "next generation"

- Input: generation, upper-limit, lower-limit
- Output: next generation of individuals



```
# Generations and fitness values will be written to this file
Result file = 'GA Results.txt'
# Creating the First Generation
def first generation(pop):
    fitness = [fitness_calculation(pop[x])_
       for x in range(len(pop))]
    sorted_fitness = sorted([[pop[x], fitness[x]]
       for x in range(len(pop))], key=lambda x: x[1])
    population = [sorted fitness[x][0]_
       for x in range(len(sorted fitness))]
    fitness = [sorted_fitness[x][1]_
       for x in range(len(sorted fitness))]
    return {'Individuals': population, 'Fitness': sorted(fitness)}
pop = population(20,8,1,0)
gen = []
gen.append(first_generation(pop))
fitness_avg = np.array([sum(gen[0]['Fitness'])/
                       len(gen[0]['Fitness'])])
fitness max = np.array([max(gen[0]['Fitness'])])
res = open(Result file, 'a')
res.write('\n'+str(gen)+'\n')
res.close()
finish = False
while finish == False:
   if max(fitness max) > 6:
       break
   if max(fitness avg) > 5:
       break
   if fitness similarity chech(fitness max, 50) == True:
        break
   gen.append(next_generation(gen[-1],1,0))
   fitness avg = np.append(fitness avg, sum(
       gen[-1]['Fitness'])/len(gen[-1]['Fitness']))
   fitness max = np.append(fitness max, max(gen[-1]['Fitness']))
   res = open(Result_file, 'a')
   res.write('\n'+str(gen[-1])+'\n')
   res.close()
```



## To be Continued....