# Metaheuristics Homework-1

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 $Homework\ Github\ URL: \underline{https://github.com/tongysmember/Metaheuristic\ hw}$ 

#### **Notes on Homework:**

- 1. Clean Code
- 2. Pseudocode
- 3. Search Feasibility
- 4. Convergence Curve

# **Question 2:**

$$f(x_1, x_2) = |x_1^2 + x_2^2 + x_1 x_2| + |sin(x_1)| + |cos(x_2)|$$
 (2)

Bound at  $[-500 \le x_i \le 500]$ . Global minimum solution at  $x^* = (0,0)$ , where  $f(x^*) = 1$ .

#### 1. Clean Code

Reference material for coding cleaner, separate every functional code into classes. And naming become reasonable. Class files as described below.

- **ALGO Folder** : Algorithm Files
  - **SA.py**: Simulated annealing algorithm
  - **GA.py**: Genetic algorithm
- **FUNC Folder**: Functional and Question.2 Files
  - GRAPH.py: Show 2D and 3D graph, made user understand convergence and iterate position easily.
  - ${\bf Objective.py}: Assigned question, it's fitness in algorithm method.$
- **HW1.py**: Main Program in homework1, just run this file by python3, it will show compute process information and results.



#### 2. Pseudocode

# 2.1 Simulate Annealing Algorighm

```
Algorithm. Simulated Annealing
-------
initinial value:
    x1, x2, fitness

while TemperatureNow > TemperatureMin:
    for iterationTimes:
        generate new x1, x2
        fitnessNew => fitness(x1, x2)
        if fitnessNew < fitness:
            Update(x1, x2)
        TemperatureNow = Temperature_decrease()

until TemperatureNow <= TemperatureMin
return Best x1, x2
```

#### 2.1 Genetic Algorithm

```
Algorithm. Genetic Algorithm
------
initinial value:
   DNA_Size, Population_Size, CrossoverRate, MutationRate, IterationTimes

for 1,2,3,... IterationTimes:
   x1, x2 => GenerateDNA()
   Population, Fitness => Selection()
    if Fitness <= Evalate:
        return Best x1, x2 Value
   Child => Crossover()
   Child_Next => Mutation()
```

# 3. Search Feasibility

#### 3.1 Assign random solution

In algorithm files, initial points and generate next-points which iterate time increase used random API method. Generated variable code as show below.

Simulated Annealing Algorithm:

```
x1 = np.random.uniform(low=self.lower_bound, high=self.upper_bound)
x2 = np.random.uniform(low=self.lower_bound, high=self.upper_bound)
```

#### Genetic Algorithm:

#### 3.2 Fit the to the end-border

Iteration result are close end-border nearly. Position of point varies very slightly. Compare with questions minimum value are very closely.

#### Simulated Annealing Algorithm:

#### Genetic Algorithm:

```
min_fitness: 1.0396797186379436
(x, y): (-0.004386063669244322, 0.2692095006528916)
min_fitness: 1.0396797186379436
(x, y): (-0.004386063669244322, 0.2692095006528916)
min_fitness: 1.004609574190532
(x, y): (-0.004386063669244322, 0.025068875425517945)
```

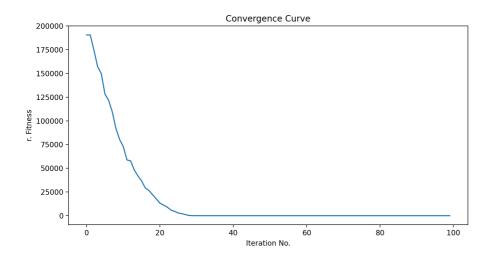
# 3.3 Overlap the move

The overlapping movement will be verified by the fitness function in the algorithm. Compared with Best\_Fitness, it is determined whether it needs to move to the next area. It is found through the 3D chart that it can move towards the minimum position.

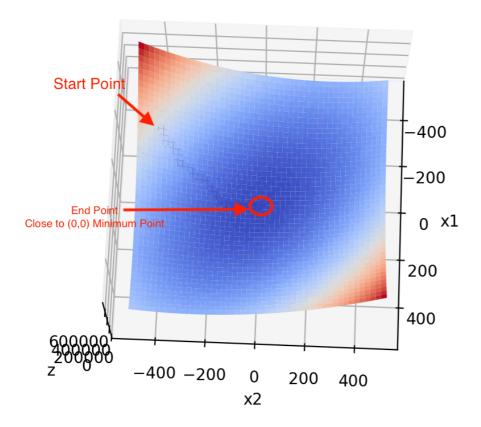
# 4. Convergence Curve

The following chart shows the convergence curve. Through the calculation results of the two algorithms, it can be seen that the fitness value continues to decrease and converge to the minimum value. The 3D graph shows that the algorithm is towards the lowest point.

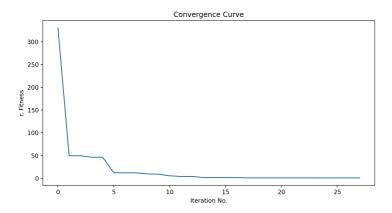
## 4.1 Simulate Annealing Algorighm



# M11015Q03\_HW1\_Q2\_Plot\_Surface



#### 4.1 Genetic Algorithm



# 5. Experimental experience

In report, I used 2 algorithms to solve minimum value. Each simulated annealing algorithm and genetic algorithm. Simulated annealing process more simple and easily implement, setting variable are few than genetic algorithm. Genetic algorithm is evolute by multiple sub process. And it needs setting more algorithm numerical value(crossover\_rate, mutation\_rate, population\_size, indivition\_size). But crossover and mutation mechanism can prevent solution stay in local optimal. It's suitable for more complex situation and need more computation.

Question 2 is a smooth surface plot. This situation isn't easy to encounter local optimal. Thus, between 2 algorithms' result are very small value. But genetic algorithm need more computation for process running.