

# MET CS 767 - Project 1

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## 1 Problem Statement

Use simulated annealing to best approximate the optimal inputs,  $(x_1, x_2, x_3)$  such that  $\forall x \in \{x_1, x_2, x_3\} (-10 \leq x \leq 10)$  and such that the following function,  $f(x_1, x_2, x_3)$ , yields the minimal output:

$$f(x_1, x_2, x_3) : x_1^2 + x_2^2 + x_3^2 \quad (1)$$

## 2 Approach

### 2.1 General Simulated Annealing

I chose to construct a general simulated annealing approach that could be used in any arbitrary problem domain, defined by the following pseudocode:

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**Require:**  $s_{init}$ , Neighbor, E, P

```
 $s \leftarrow s_{init}$ 
 $e \leftarrow E(s)$ 
 $t \leftarrow 1.0$ 
 $t_{min} \leftarrow 0.0001$ 
 $\alpha \leftarrow 0.9$ 
 $k \leftarrow 1000$ 
while  $t \geq t_{min}$  do
  for all  $0 \leq i < k$  do
     $s' \leftarrow \text{Neighbor}(s)$ 
     $e' \leftarrow E(s')$ 
     $p_{accept} \leftarrow P(e, e', t)$ 
     $p_{rand} \leftarrow \text{Random}(0, 1)$ 
    if  $p_{accept} > p_{rand}$  then
       $s \leftarrow s'$ 
       $e \leftarrow e'$ 
    end if
  end for
   $t \leftarrow t * \alpha$ 
end while
return  $(s, e, t)$ 
```

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The inputs and variables for the above formula are defined as the following:

- $s_{init}$ : Initial state within the problem domain. This is to be chosen at random.
- $\text{Neighbor}(s)$ : Function that accepts a state,  $s$ , as input and outputs a randomly chosen local neighbor,  $s'$ , within the problem domain. It should randomly select a neighbor that is considered local to  $s$ .
- $E(s)$ : Function that accepts a state,  $s$ , as input and outputs the state's energy,  $e$ , with regards to the problem space. The energy should be the output of the problem function  $f$  with  $s$  as the input.
- $P(e, e', t)$  Function that accepts the energy,  $e$ , of current state,  $s$ , the potential energy,  $e'$ , of new state,  $s'$ , and the annealing temperature,  $t$ . It outputs an acceptable probability,  $p_{accept}$ , such that the new state,  $s'$ , is accepted as the current state,  $s$ , if  $p_{accept} > \text{Random}(0, 1)$ . This function should return  $p_{accept} \leftarrow 1$  if  $e' < e$ . Otherwise,  $p_{accept} \leftarrow \exp(e - e')/t$ .

- $t$ : The progressively cooling temperature of the annealing system. Starts at a temperature of 1 unit and eventually reduced to  $t_{min}$ . For each iteration,  $t$  is reduced in the following way:  $t \leftarrow t * \alpha$ .
- $t_{min}$ : The temperature to which  $t$  should reach before the system exists
- $\alpha$ : The rate of temperature decrease of  $t$  for each annealing iteration.
- $k$ : The number of steps for each temperature iteration. During each step, a new potential state,  $s'$ , and potential energy,  $e'$ , are generated and conditionally set to  $s$  and  $e$ , respectively given an acceptance probability  $p_{accept}$ .

## 2.2 Problem-Specific Simulated Annealing

It is the responsibility of the problem space to define  $s_{init}$ , Neighbor, E, P as defined above. For the problem at hand, a state  $s$  is defined as a tuple of  $(x_1, x_2, x_3)$  values.  $s_{init}$  is defined as randomly chosen:

$$s_{init} \leftarrow (\text{Random}(-10, 10), \text{Random}(-10, 10), \text{Random}(-10, 10)) \quad (2)$$

Futhermore, Neighbor is defined as a function that takes state,  $s = (x_1, x_2, x_3)$ , as input and outputs a local neighbor,  $s'$ , in which a randomly chosen  $x_i$  for  $i \in \{1, 2, 3\}$  is updated with a random value,  $\text{Random}(-10, 10)$ :

$$\text{Neighbor}(s) : s' \leftarrow s; s'_{i \leftarrow \text{Random}(1,2,3)} \leftarrow \text{Random}(-10, 10) \quad (3)$$

Furthermore, E is defined as a function that takes a state,  $s = (x_1, x_2, x_3)$ , as input and outputs an energy,  $e$ , set to the result of problem function,  $f$ , as such:

$$E(s = (x_1, x_2, x_3)) : e \leftarrow f(x_1, x_2, x_3) \quad (4)$$

Furthermore, P is an acceptance probability function that takes as input a current energy,  $e$ , a potential energy,  $e'$ , and temperature,  $t$ . It then outputs a probability,  $p_{accept}$  such that:

$$P(e, e', t) : p_{accept} \leftarrow (1.0 \text{ If } (e' < e) \text{ Otherwise } \exp((e - e')/t)) \quad (5)$$

## 3 Implementation and Project Layout

The entire solution is implemented in JavaScript with a HTML frontend. Specifically, the algorithm is broken up into two files:

- `js/SimulatedAnneal.js`: General simulated annealing algorithm that takes  $s_{init}$ , Neighbor, E, P as input.
- `js/MinSumOfSquares.js`: Problem specific simulated annealing algorithm that implements  $s_{init}$ , Neighbor, E, P, and leverages `js/SimulatedAnneal.js` for the general simulated annealing heuristic.
- `js/PRG.js`: A custom pseudo random generator that accepts a seed tuple of integerts,  $(m, c)$ . This was necessary because pseduorandom number generators vary from browser to browser.

In addition, the project includes a front end which allows a user to configure a seed,  $(m, c)$ , for the PRG, as well as values:  $t, t_{min}, \alpha, k$ , for the the simulated annealing algorithm. The frontend outputs a list of the states,  $s_i \leftarrow (x_1, x_2, x_3)$  and energies,  $e$ , for each temperature,  $t$ , annealing iteration,  $i$ .

The front end also outputs an animation of the problem space with each state,  $s_i$  for each annealing iteration. The animation is a cube in which each dimation is different value,  $x_i$  for  $i \in \{1, 2, 3\}$ . Over time, states  $s_i$ , should converge towards the middle of the box where  $(x_1, x_2, x_3) = (0, 0, 0)$ .

This functionality is implemented in the following files:

- `img/axis.png`: Represents each of the dimentions of the animation -  $(x_1, x_2, x_3)$ .
- `js/Chart.js`: Generates the animation using HTML's canvas element.

- `style/normalize.css`: Standardize the css styles across all browsers (i.e. fonts, sizes, padding, etc ...)
- `style/style.css`: Styles specific to the project frontend in `index.html`.
- `index.html`: Frontend HTML code to run algorithm and resulting animation.

### 3.1 Execution

Set up a simple python server to serve up `index.html` and corresponding `js/*`, `style/*`, `img/*` resources.

```
$ cd /path/to/project/
$ python -m SimpleHTTPServer
```

This should display something like: "Serving HTTP on 0.0.0.0 port 8000 ...". At that point, go to `http://localhost:8000` to view the frontend. (Note that this is not required if viewing the frontend with Google Chrome. In such, just open `index.html` in the Chrome browser: e.g. `file:///<path-to-project>/index.html`)

## 4 Inputs and Solution

### 4.1 Inputs

The following inputs were used. Each are the default within the configuration on the frontend:

- $t$ : 1.0
- $t_{min}$ : 0.0001
- $\alpha$ : 0.9
- $k$ : 100
- $seed_m$ : 2319871
- $seed_c$ : 93717033

### 4.2 Solution

The resulting, best-effort, optimal solution state was:  $(x_1, x_2, x_3) = (-0.0902, -0.138, -0.127)$ . The solution was generated over 88 iterations of annealing to reach a temperature of 0.000104. The resulting minimum value of  $f$  was:  $\min(f(x_1, x_2, x_3)) = 0.0432$ .

These results may be reproduced by simply using the defaults on the frontend web application.

## 5 Resources

- [https://en.wikipedia.org/wiki/Simulated\\_annealing](https://en.wikipedia.org/wiki/Simulated_annealing)