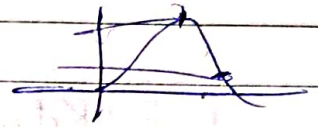


Simulated annealing



Step 1: Set $T = T_{\text{initial}}$

Step 2: Set $x_{\text{best}} = x_{\text{current}}$

// Initialize the best solution

Step 3: while $T > T_{\text{min}}$

// Initialize T_{min} in beginning

Step 4: for $i = 1$ to max-iterations

// Initialize max-iterations

Step 5: Generate a new solution

x_{new} by slightly modifying x_{current}

Step 6: Calculate $\Delta E = f(x_{\text{new}}) - f(x_{\text{current}})$

Step 7: If $\Delta E < 0$ (x_{new} is better)

Step 8: Accept $x_{\text{new}} : x_{\text{current}} = x_{\text{new}}$

else:

Accept x_{new} with probability

$$P = \exp(-\Delta E / T)$$

\approx

- if accepted, set

$$x_{\text{curr}} = x_{\text{new}}$$

9.

If $f(x_{\text{curr}}) < f(x_{\text{best}})$

update $x_{\text{best}} = \text{current}$

10. Reduce the temperature

$$T = T \times \alpha.$$

11. Return x_{best} as the final solution.

Python code:

```
import random
import math
```

```
def objective_func(x):
    return  $x^{x+2}$ 
```

```
def sa(initial_sol, init_temp, cool_rate, max_it):
    curr_sol = initial_sol
    curr_temp = init_temp
    best_sol = curr_sol
    best_val = objective_func(best_sol)
```

```
    curr_val = best_val
```

```
    for it in range(max_it):
```

```
        new_sol = curr_sol + rand.uniform(-1, 1)
        new_val = objective_func(new_sol)
```

```
        delta_val = new_val - curr_val
```

```
        if delta_val < 0 or rand.random() < math.exp(-delta_val / curr_val):
```


curr_sol = new_sol
 current_val = new_val

if new_val < best_val:
 best_sol = new_sol
 best_val = new_val

curr_t * = cool_rate

return best_sol, best_val

initial_sol = random.uniform(-10, 10)
 initial_te = 1000
 cool_rate = 0.99
 max_it = 1000

best_sol, best_val = simulate_an(initial_sol, initial_te,
 cool_rate, max_it)

print(f"The best x : {best_sol}, It's corresponding
 $f(x)$: {best_val}")

output:

The best x : -0.0008206, It's corresponding
 $f(x)$: 6.7349e-07

Am
 22/10/21