# COL333 Assignment 1

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Optimization Problem using Greedy Hill Climbing with Random Restart

#### **Initialization:**

- Initiate the optimization process using Greedy Hill Climbing with Random Restart.
- Randomly select a permutation of locations, denoted as "init\_state."
- The first z locations in "init\_state" correspond to the zones and are allocated sequentially.

#### **Cost Calculation:**

- Compute the cost of the initial state using a complexity function with  $O(z^2)$  complexity.

#### Pair Generation:

- Define a vector named "all\_pairs" comprising pairs representing all possible mappings of zones and locations.
- The size of "all-pairs" is  $(z \times l)$ , where z is the number of zones, and l is the number of locations.

## Hill Climbing Step:

- In each iteration of the hill climbing step:
  - 1. Randomly generate a number, r, between 1 and  $l \times z$ .
  - 2. Use r to select a pair (i, j) from the "all\_pairs" vector.
  - 3. Swap the locations at indices i and j in the "init\_state."

### **Cost Evaluation:**

- Calculate the partial contribution to the cost before and after the location swap in O(z) complexity

### Improvement Check:

- Check for an improvement in the total cost after the swap.
- If an improvement is observed, restart the iterative step.
- If no improvement occurs after  $l \times z$  swaps, proceed to the next step.

#### **Random Restart:**

- In case of no improvement after  $l \times z$  swaps, initiate a random restart.
- Generate a new permutation of locations in "init\_state."
- Recalculate the cost for the newly generated state.

#### **Iteration Control:**

 Continue the iterative process, incorporating hill climbing steps and random restarts, until the allocated time is exhausted.

# Another Approach:

In this variant, two random indices i and j (bounded by  $1 \le i \le z$  and  $1 \le j \le l$ ) are generated. The locations at these indices in "init\_state" are swapped, and the partial cost impact is assessed. If no cost improvement is observed after a specified number of swaps, the entire process is restarted. This method introduces a dynamic element through random swapping and employs a restart mechanism for optimization.

# Probabilistic Swapping:

- Incorporate a probabilistic approach in the swapping process:
- With probability  $\frac{l-z}{l}$ , swap unallocated and allocated locations.
- With probability  $\frac{z}{l}$ , swap allocated locations.

#### **Conclusion:**

- In our exploration, it has been determined that the performance of the algorithm improves as the number of climbing steps increases.
- This approach introduces randomness through random swapping, providing a dynamic element to the optimization process.
- The probabilistic approach further diversifies the search space, contributing to the overall efficacy of the algorithm.