

Assignment (1)

Name : Samiha Akter Maisha

Roll no : 01122B1010

Section : (D)

Course : Artificial Intelligence

Question no(1)

* Hill Climbing :

Merits : Easy to implement and fast conjunction

Demerits : Easily get stuck in local optima and there is no backtracking.

Handling Local Minima : Easily get stuck in local minima.

Global Optimality : Unlikely to find global optimum.

Simulated Annealing :

Merits : It can escape local minima using random exploration, balance exploration and exploitation.

Demerits : Slower conjunction and requires tuning of temperature schedule.

Handling Local minima : It's good because probabilistically escape local minima.

Global Optimality : Moderate may reach global optimum with good tuning.

Beam Search :

Merits : Efficient with large state space and uses multiple candidates.

Demerits : May converge to suboptimal solution if beam width is small and not guaranteed to explore all paths.

Handling Local Minima : It's Medium and depends on the beam width and heuristic quality.

Global Optimality : Moderate better than hill climbing but not guaranteed.

Genetic Algorithm:

Merits : It explores large search space, maintains diversity, high adaptability.

Demerits : It explores large search space, maintains diversity, high adaptability.

Demerits : Requires careful parameters tuning and slow for every big problems.

Handling Local Minima : It's very good and maintains multiple diverse solution

and uses crossover.

Global Optimality : High and high chance of finding global optimum.

2 no question

Hill Climbing:

Efficiency: It's very fast and low memory usages.

Complexity: $O(b^d)$. Here b is branching factor and d is depth.

Scalability: Poor and doesn't scale well. and sometimes performance drops.

Simulated Annealing:

Efficiency: Slower due to probabilistic decisions.

Complexity: $O(n^2) + O(n^3)$ depends on cooling schedule

Scalability: Moderate. It handles larger spaces better than hill climbing.

Beam Search:

Efficiency: Moderate efficiency and parallelizable.

Complexity: $O(k \times b^d) - k$ = beam width.

Scalability: Good because it can increase beam width to handle bigger problems.

Genetic Algorithm:

Efficiency: It's computationally expensive

Complexity: $O(g \times p \times n) - g$ = generation, p = population size, n = problem size.

Scalability: Very good and ideal for large, complex problems.

3no ans

Based on my analysis the most suitable algorithm is Genetic Algorithm.

Justifying my recommendation below:

- (1) It handles large solution spaces efficiently.
- (2) Maintains diversity of routes using populations.
- (3) Escapes local minima using mutation and crossover.
- (4) Can incorporate constraints like delivery time window, traffic etc.

Optimal Combination:

Use simulating annealing to refine the best solution from the Genetic Algorithm to further improve local optimization.