

Assignment (1)

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Section : (D)

Course : Artificial Intelligence

* Hill Climbing:

Merits: Easy to implement and fast conjunction

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

Handling Local Minima: It is poor cause gets 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 coverage 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 % 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.

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.