

Local Search

✱✱

Uninformed Search } Uses backtracking because
Informed Search } path matters there, perfect
goal matters.

✱✱ Uninformed & Informed searches are very expensive in large scale graphs/infinite graphs.

✱✱ There are problems where we are not concerned with path. Only the solution itself matters.

↳ Not concerned with perfect solution.

↳ So backtracking & path এর কথা উঠে
Node সতে হবে, এই ডিভিশনাল বাদ দিয়ে যেই
algorithm খুঁজ/ search algorithm খুঁজ
discuss হবে হয়, তাদের local search হবে।

✱✱ Advantages of local search

① time কম লাগে

② Memory কম লাগে

③ Large / infinite graph এ খুঁজ করা সহজ।

Example: Hill climbing algorithm, simulated annealing.

** Approaches of local Search:

↳ Keep a "current" state & tries to improve it.

↳ Starts with an ~~empty~~ initial guess of solution and gradually improving it until it is the optimal one.

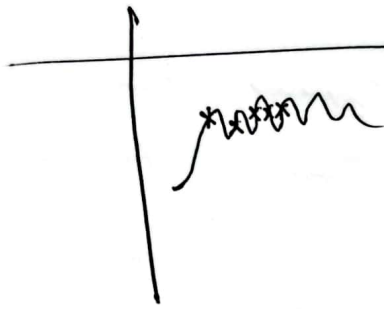
** Disadvantages of local Search:

① Local Maxima: There can be multiple local maxima. So depending on starting point it can reach a local maxima and terminate the algorithm. In this case it returns top point but not the best one.

② Plateaus: The space has a broad flat region. So the algorithm can not move forward or backward because the next or previous point is not leading towards anything, nor the optimal point nor the sub-optimal point. So it gets stuck.



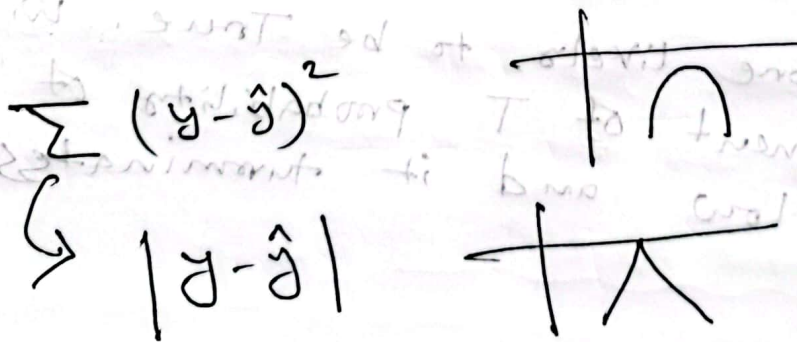
- ③ Ridges: Multiple local maximas are closed to each other. So a small step can lead to random sub optimal points.



✳️ Solution:

- ① Random restart: After reaching a top randomly restart to find the global maxima.

- ② Problem reformulation: So, we can reformulate our problem space to avoid ~~unavoidable~~ situations.



③ Simulated Annealing:

$C = C_{int}$ # initializing current state

for $T = T_{max}$ to T_{min} :

$$E_c = E(C)$$

$$N = \text{Next}(C)$$

$$E_N = E(N)$$



$$\Delta E = E_N - E_c$$

if $\Delta E > 0$:

$$C = N$$

else if $(e^{\Delta E/T} > \text{rand}(0,1))$:

$$C = N$$

End

when T is large $e^{\Delta E/T}$ is large, so it is more likely to be True. With the decrement of T probability of 'True' becomes low and it terminates.

Informed Search

Uniform Cost Search (informed search)

→ Path cost from 'slant node' to 'n'

- Evaluation function, $f(n) = g(n)$

Greedy Best First Search (GBFS)

→ heuristic

- Evaluation function, $f(n) = h(n)$

☐ GIBFS Optimal?

- Not optimal. Because let True path cost considers ∞ , approximate ∞ heuristic considers ∞ , so approximate ∞ soln path cost.

Q4 GBFS Complete?

- Tree: Not complete. visited / expanded track
 রাখা হয় না, so ওরকি Node যার যার push
 হয়, সেটা ও পড়ে যায়।

In case of both finite and infinite tree it is not complete.

- Graph: Complete for finite graph. loop prevent ~~for for~~ but for infinite it's not complete.

Time and Space Complexity?

- $O(b^m)$ (Same as bfs, dfs) Practical implementation is heuristic rather quality or rather Time & Space Complexity rather exact.

A* Search

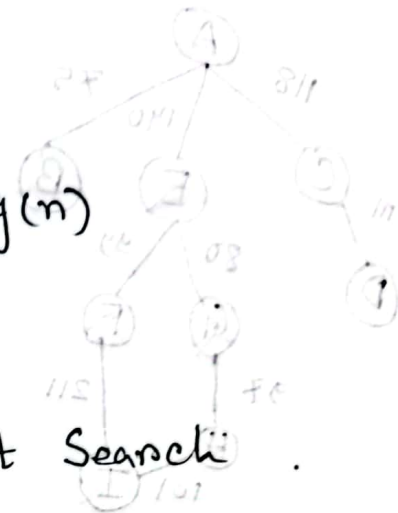
- Evaluation function, $f(n) = h(n) + g(n)$
- when $h(n) = 0$:

A* Search = Uniformed Cost Search

when $g(n) = 0$:

A* Search = GBFS

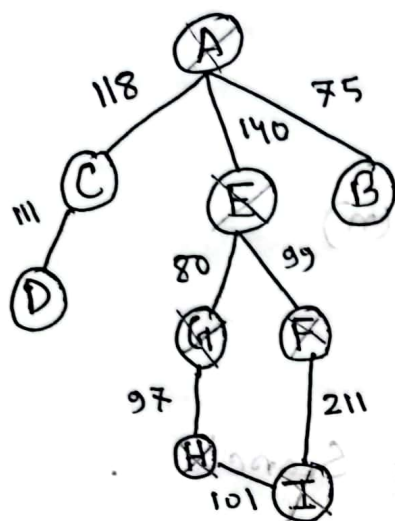
Node	Parent
A	-
B	A
C	A
D	B
E	B
F	E
G	E
H	F
I	G



Node	Parent
A	-
B	A
C	A
D	B
E	B
F	E
G	E
H	F
I	G

Node	Parent
A	-
B	A
C	A
D	B
E	B
F	E
G	E
H	F
I	G

A* Search Simulation



State	$h(n)$
A	366
B	374
C	329
D	244
E	253
F	178
G	193
H	98
I	0

Solⁿ:

A ³⁶⁶	C ¹¹⁸ , E ¹⁴⁰ , B ⁷⁵
E ³⁹³	C ⁴⁴⁷ , B ⁴⁴⁹ , G ²²⁰ , F ²³⁹
G ⁴¹³	C ⁴⁴⁷ , B ⁴⁴⁹ , F ⁴¹⁷ , H ³¹⁷
H ⁴¹⁵	C ⁴⁴⁷ , B ⁴⁴⁹ , F ⁴¹⁷ , I ⁴¹⁸
F ⁴¹⁷	C ⁴⁴⁷ , B ⁴⁴⁹ , I ⁴¹⁸ , I ⁴⁵⁰
I ¹¹⁸	

Path: A³⁶⁶ → E³⁹³ → G⁴¹³ → H⁴¹⁵ → I¹¹⁸

Cost:

Parent	Node	Distance from start, $g(n)$
-	A ³⁶⁶	0
A ³⁶⁶	C ⁴⁴⁷	0 + 118 = 118
A ³⁶⁶	E ³⁹³	0 + 140 = 140
A ³⁶⁶	B ⁴⁴⁹	0 + 75 = 75
E ³⁹³	G ⁴¹³	140 + 80 = 220
E ³⁹³	F ⁴¹⁷	140 + 99 = 239
G ⁴¹³	H ⁴¹⁵	220 + 97 = 317
H ⁴¹⁵	I ⁴¹⁸	317 + 101 = 418
F ⁴¹⁷	I ⁴⁵⁰	239 + 211 = 450
I ¹¹⁸		

** Admissibility:

$$\bullet h(n) \leq C(n, \text{Goal})$$

→ A graph's heuristic is admissible when this condition is true for all the nodes.

If one Node breaks this condition, the graph's heuristic is not admissible.

and when the condition breaks we say "h" is overestimated.

• When heuristic is not admissible, it gives sub-optimal path.

Example:

E

$$\rightarrow h(E) = 253$$

$$\rightarrow C(E, \text{Goal}) = \text{lowest}(278, 310) = 278$$

$$253 \leq 278 \quad \checkmark$$



A* Search optimal?

- efficient path that for $(n)N$.

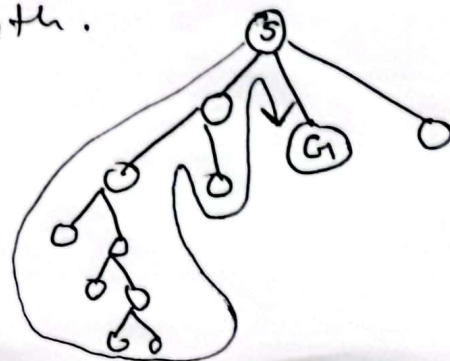
- if heuristic is Admissible:
Optimal

else:

Not Optimal

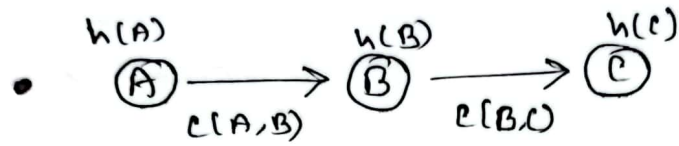
When $h(n)$ is "optimal" but $h(n)$ is not "consistent" then it takes more time to find the optimal path.

When $h(n)$ is "Optimal" but "not consistent". It surely gives optimal path but take more time. Roams around unwanted paths. then finds optimal path.



Consistency:

- $h(n) \leq c(n, n') + h(n')$



$$\begin{aligned} \hookrightarrow h(A) &\leq c(A, B) + h(B) \\ \hookrightarrow h(B) &\leq c(B, C) + h(C) \end{aligned}$$

$\Rightarrow h(n)$ is "consistent" is the condition is true for all the nodes. (except Goal)

$\Rightarrow h(n)$ can be Admissible but not consistent

\hookrightarrow less time & space taken

A* Search Complete? (\leftarrow loop & node फिर, unwanted nodes & पुनरावृत्ति node फिर)

- Graph Search: if $h(n)$ Admissible:

Complete

else:

Not Complete

- Tree Search: if $h(n)$ is Admissible & Consistent:

Complete

else:

Not Complete