



Artificial and Computational Intelligence

AIMLCLZG557

Contributors & Designers of document content : Cluster Course Faculty Team

M2 : : Problem Solving Agent using Search



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Artificial and Computational Intelligence

Disclaimer and Acknowledgement



- Few content for these slides may have been obtained from prescribed books and various other source on the Internet
- I hereby acknowledge all the contributors for their material and inputs and gratefully acknowledge people others who made their course materials freely available online.
- .I have provided source information wherever necessary
- This is not a full fledged reading materials. Students are requested to refer to the textbook w.r.t detailed content of the presentation deck that is expected to be shared over e-learning portal - taxilla.
- I have added and modified the content to suit the requirements of the class dynamics & live session's lecture delivery flow for presentation
- **Slide Source / Preparation / Review:**
 - From BITS Pilani WILP: Prof.Raja vadhana, Prof. Indumathi, Prof.Sangeetha
 - From BITS Oncampus & External : Mr.Santosh GSK

Course Plan



M1 Introduction to AI

M2 Problem Solving Agent using Search

M3 Game Playing

M4 Knowledge Representation using Logics

M5 Probabilistic Representation and Reasoning

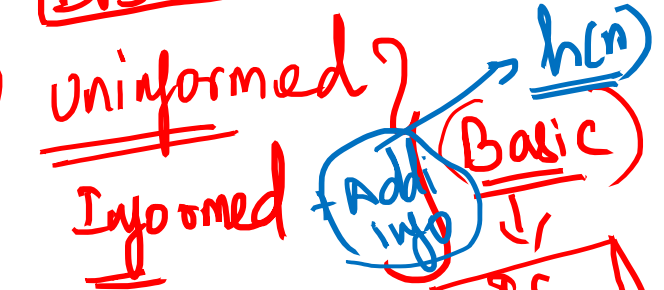
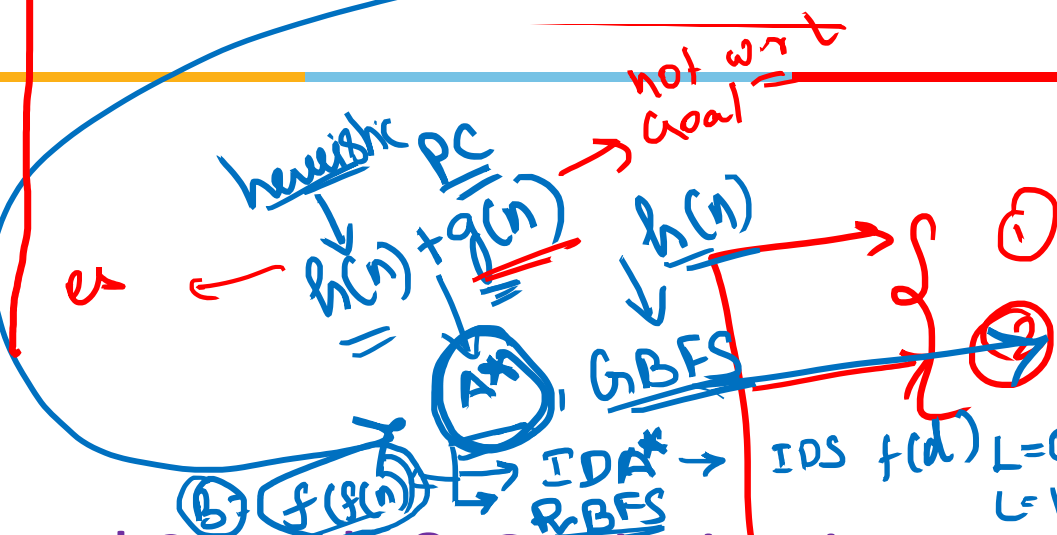
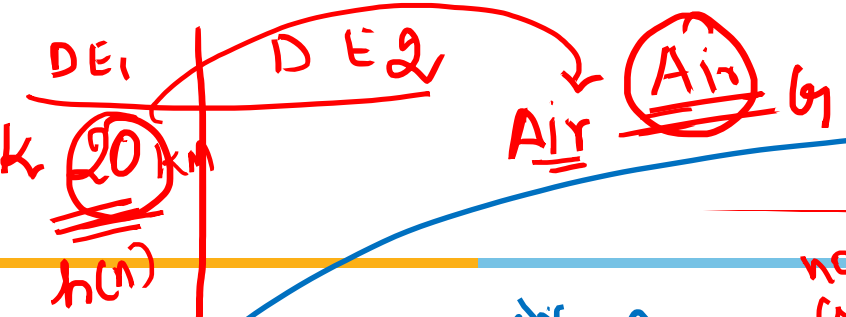
M6 Reasoning over time, Reinforcement Learning

M7 Ethics in AI

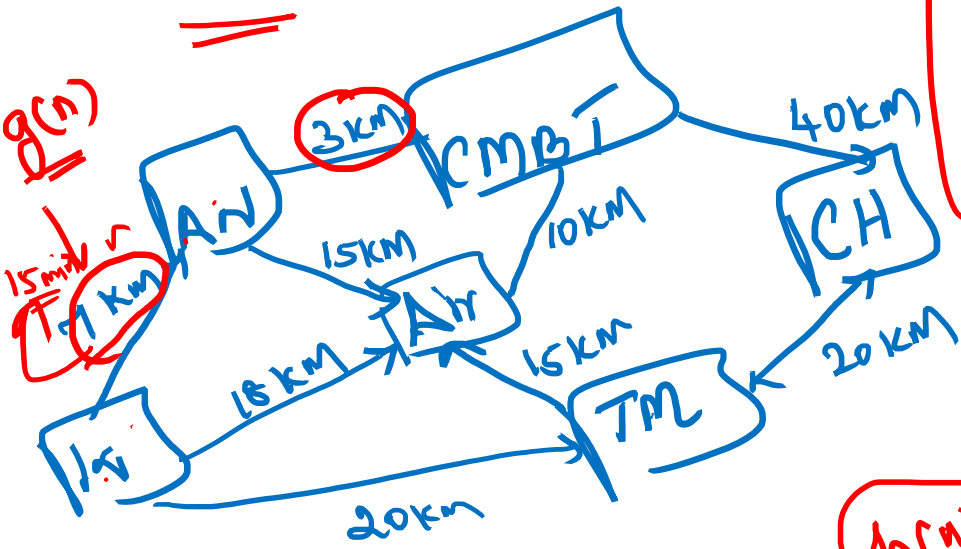
Learning Objective

At the end of this class , students Should be able to:

1. Design fitness function for a problem
2. Construct a search tree
3. Apply appropriate local search and show the working of algorithm at least for first 2 iterations with atleast four next level successor generation(if search tree is large)
4. Design and show Genetic Algorithm steps for a given problem



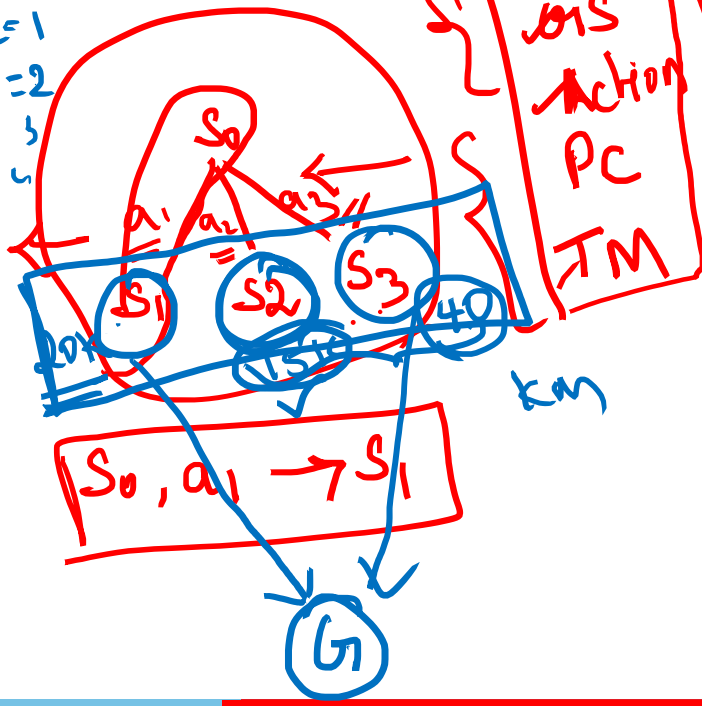
Local Search & Optimization



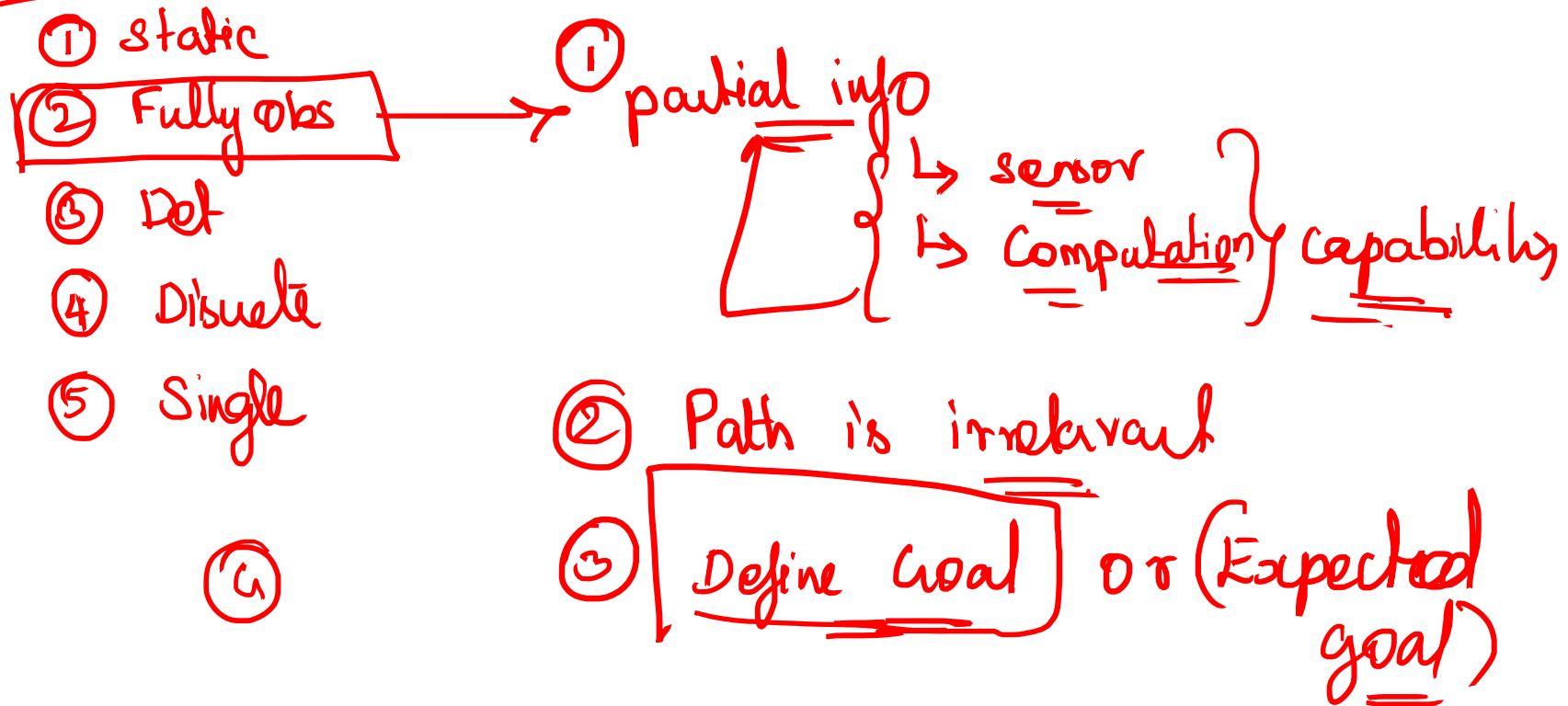
Search tree

TM

h(n)



Envi



○○○

Chess board →

← "Greedy choice" → 'LOCAL Search'

Optimization Problem

Goal : Navigate through a state space for a given problem such that an optimal solution can be found

Objective : Minimize or Maximize the objective evaluation function value

Scope : Local

best Successor

Objective Function : (Fitness Value) evaluates the goodness of current solution

$h(n)$

Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found

① Single Instance Based

✓ Hill Climbing → RRHC, SHC

~~Simulated Annealing~~

✓ Local Beam Search

~~Tabu Search~~

② Multiple Instance Based

✓ Genetic Algorithm

✓ Particle Swarm Optimization

✓ Ant Colony Optimization

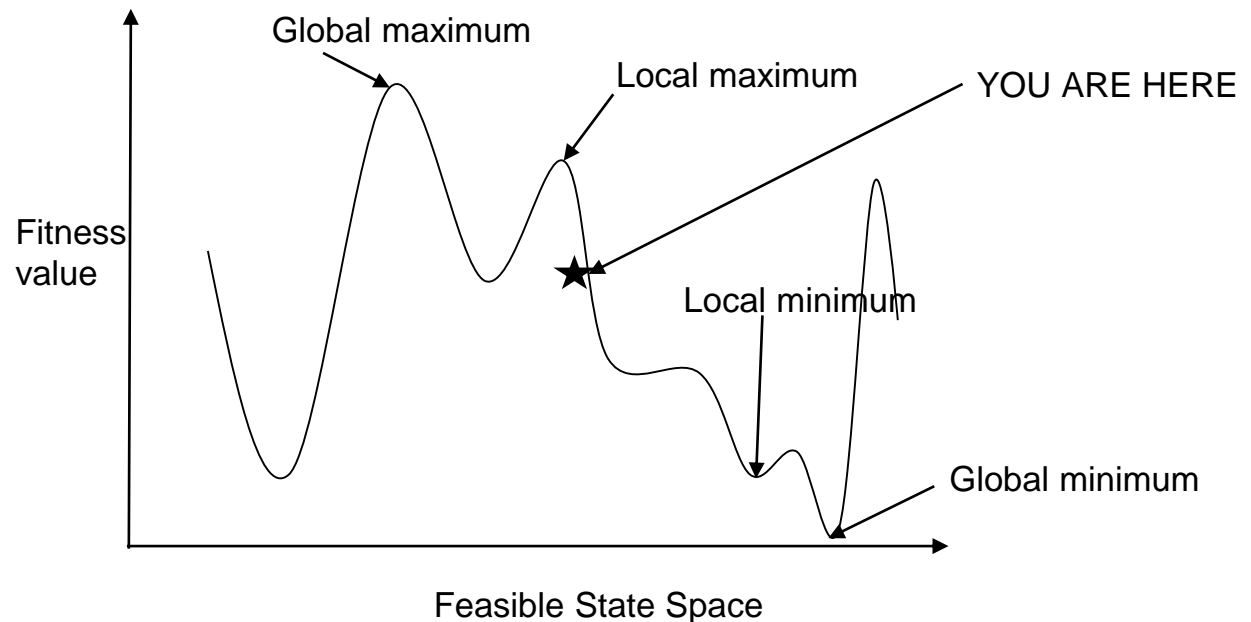
Terminology

Local Search : Search in the state-space in the **neighbourhood** of current position until an optimal solution is found

Algorithms:

- Choice of Neighbor
- Looping Condition
- Termination Condition

2	5	3	2
♠	6	♠	♠
3	5	4	2
4	♠	4	2



Local Search

Empty

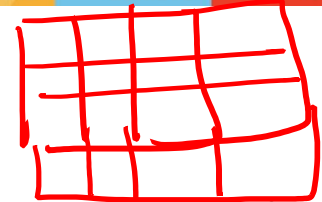
Local

completely filled BC

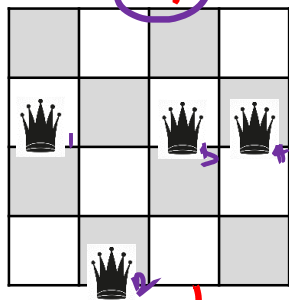
PF BC

Empty

Terminology



Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found



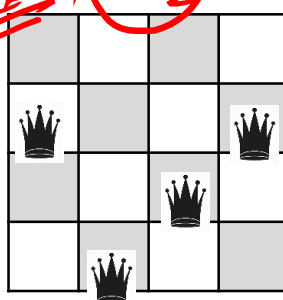
a_1, b_2
 a_1, b_3
 a_1, b_4
 a_2, b_3
 a_2, b_4

Feasible State/Solution

Fitness Value:

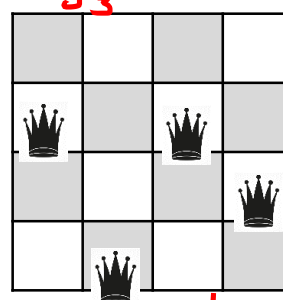
$h(n) = 4$

$h(n) = \text{No. of Conflicting pairs of queens}$ min

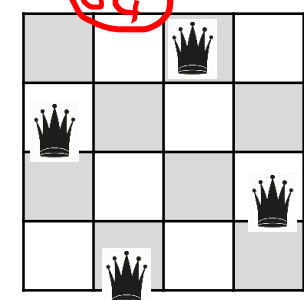


Neighboring States

$h(n) = 4$



$h(n) = 2$



Optimal Solution

$h(n) = 0$

$h(n) = 2$

$h(n) = 2$

$h(n) = 4$

$h(n) = 6$

$h(n) = \text{No. of Non-Conflicting pairs of queens.}$ max

Local Search

Empty

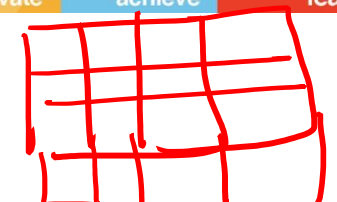
Local

completely filled BC

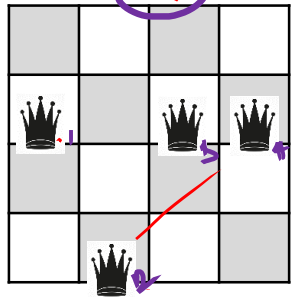
PF BC

Empty

Terminology



Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found



6 pairs

$Q_1 Q_2$ (NC)

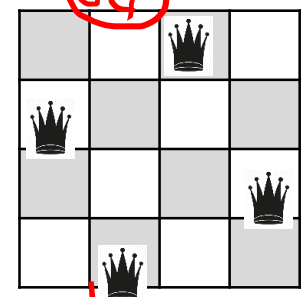
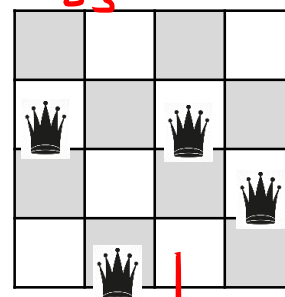
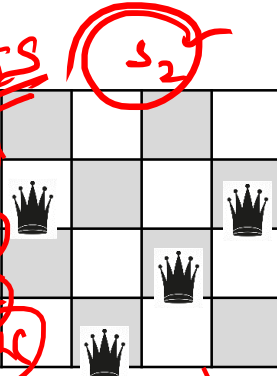
$Q_1 Q_3$ (C)

$Q_1 Q_4$ (C)

$Q_2 Q_3$ (C)

$Q_2 Q_4$ (C)

$Q_3 Q_4$ (C)



Feasible State/Solution

Neighboring States

Optimal Solution

Fitness Value:

$h(n) = 4$

$h(n) = 4$

$h(n) = 2$

$h(n) = 0$

$h(n) = \text{No. of Conflicting pairs of queens}$

$h(n) = 2$

$h(n) = 2$

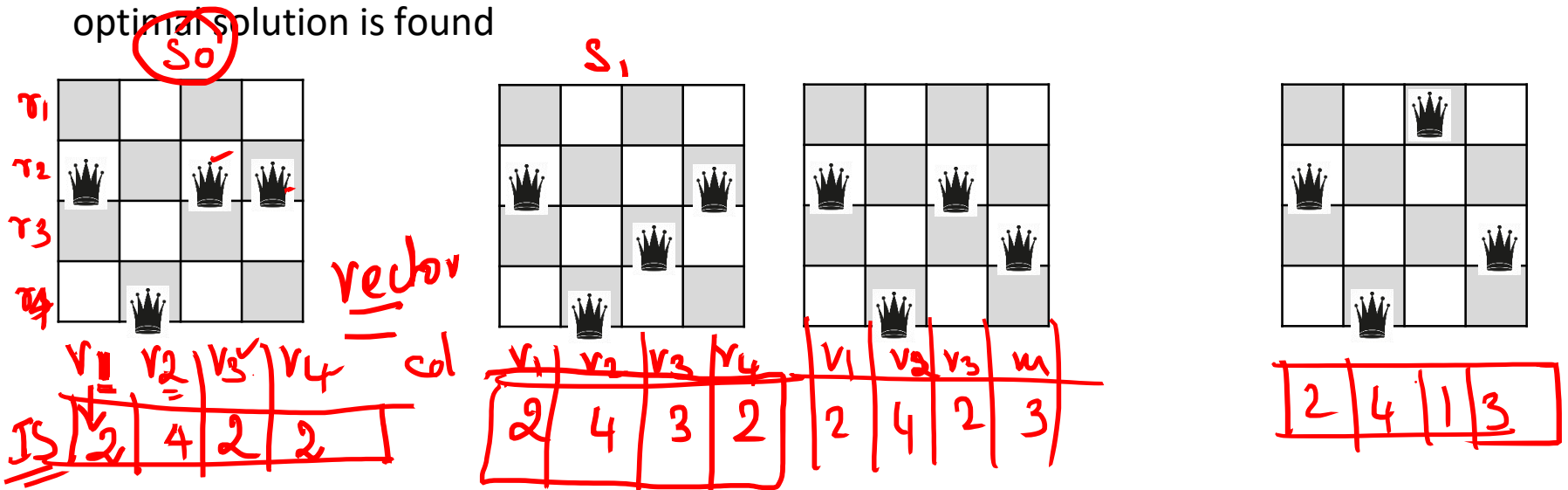
$h(n) = 4$

$h(n) = 6$

$h(n) = \text{No. of Non-Conflicting pairs of queens.}$

Terminology

Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found



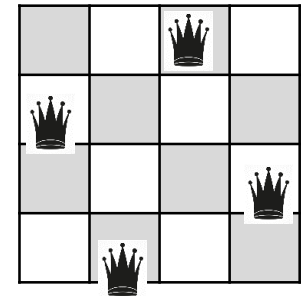
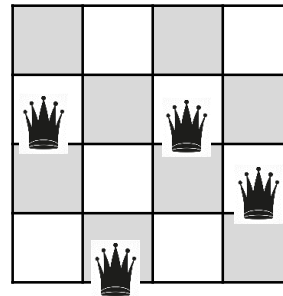
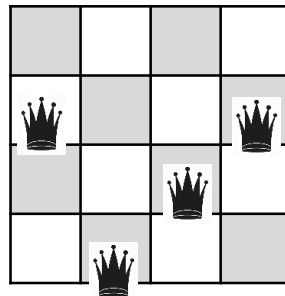
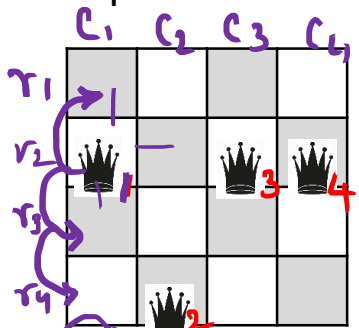
Local Search



V_1 → position of var v on these rows
 v in the 1st col positioned at row 2

Terminology

Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found



So

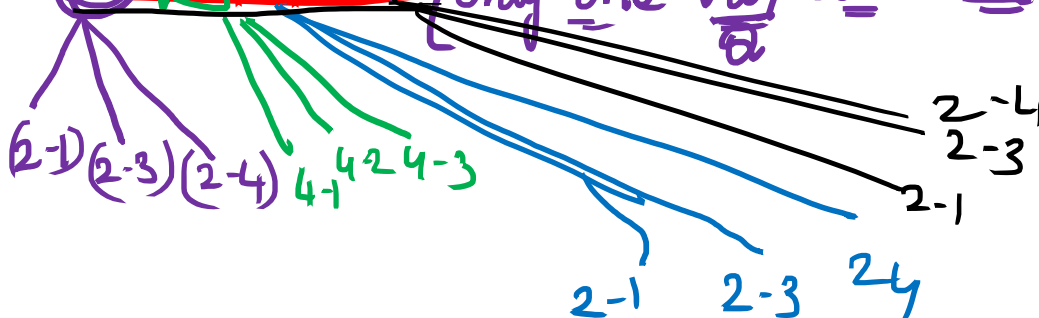
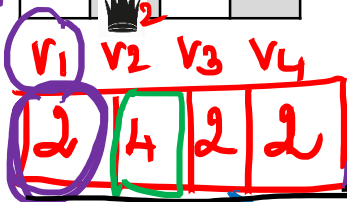
LS change
 [only one var at a time]

1 NN

1 Neighborhood region expansion

9

2 NN



12 Success

Local Search

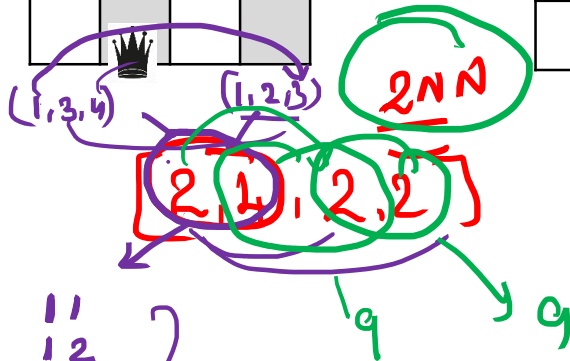
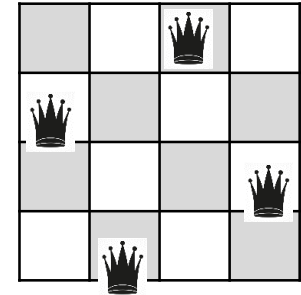
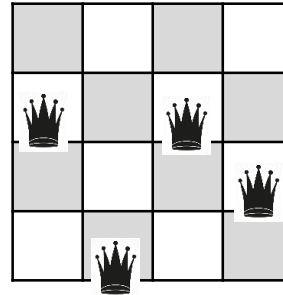
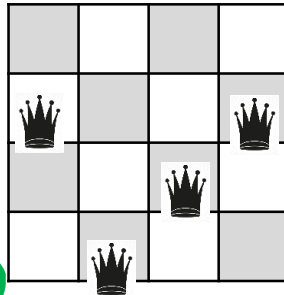
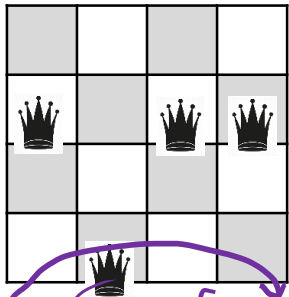


3NN → ↑ed
4NN → ↑ed

Terminology



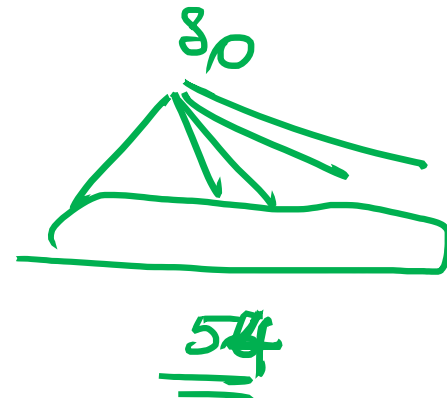
Local Search : Search in the state-space in the neighbourhood of current position until an optimal solution is found

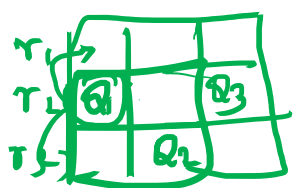


1	1
1	2
1	3
<hr/>	
3	1
3	2
3	3
<hr/>	
4	1
4	2
4	3

$v_1 v_2 \rightarrow 9$
 $v_1 v_3 \rightarrow 9$
 $v_1 v_4 \rightarrow 9$
 $v_2 v_3 \rightarrow 9$
 $v_2 v_4 \rightarrow 9$
 $v_3 v_4 \rightarrow 9$

6×9
54





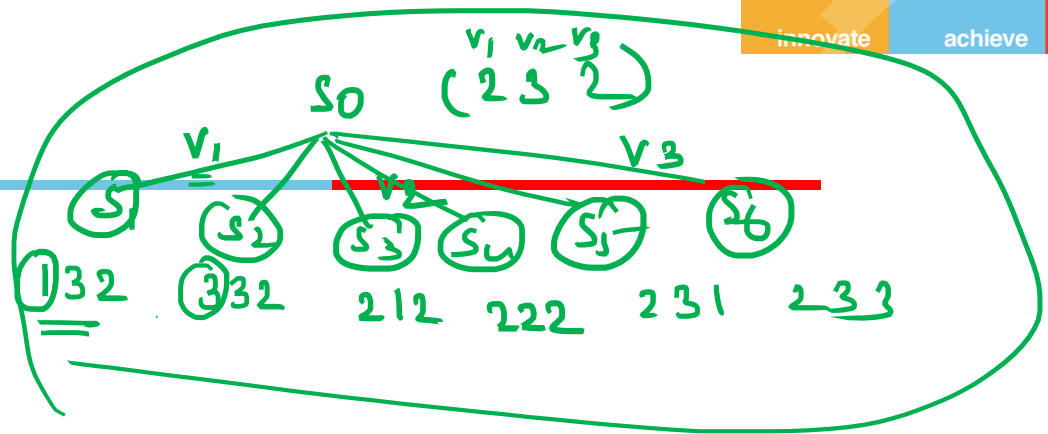
$v_1 \ v_2 \ v_3$

2 3 2



1NN

2



6

2NN

2 (3) 2

2 3

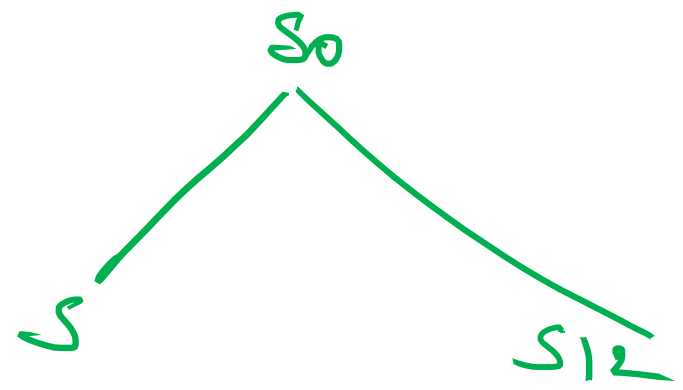
$v_1 \ v_2 \rightarrow 4$

$v_1 \ v_3 \rightarrow 4 \Rightarrow \underline{\underline{12,8}}$

$v_2 \ v_3 \rightarrow 4$

Hill Climbing

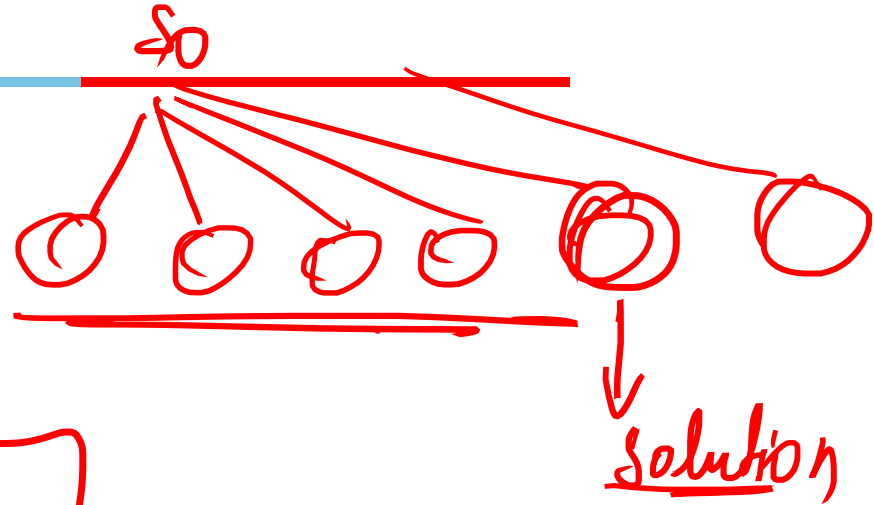
2 3 2



1, 1
1, 2
3, 1
3, 2

$(1,3)$ $(1,2)$ $(1,3)$
2 3 2

1 1 1

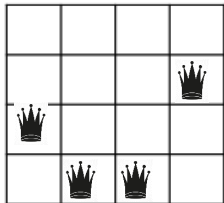


1 NN

Hill Climbing

Hill Climbing

1. Select a random state
2. Evaluate the fitness scores for all the successors of the state
3. Select the next state based on the highest fitness
4. Repeat from Step 2



3	4	4	2	3
---	---	---	---	---

function HILL-CLIMBING(*problem*) **returns** a state that is a local maximum

current \leftarrow MAKE-NODE(*problem*.INITIAL-STATE)

loop do

neighbor \leftarrow a highest-valued successor of *current*

if *neighbor*.VALUE \leq *current*.VALUE **then return** *current*.STATE

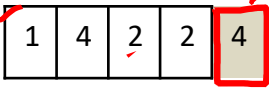
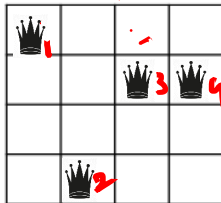
current \leftarrow *neighbor*

Hill Climbing

1. Select a random state
2. Evaluate the fitness scores for all the successors of the state

$h(n)$ = No. of non-conflicting pairs of queens in the board.

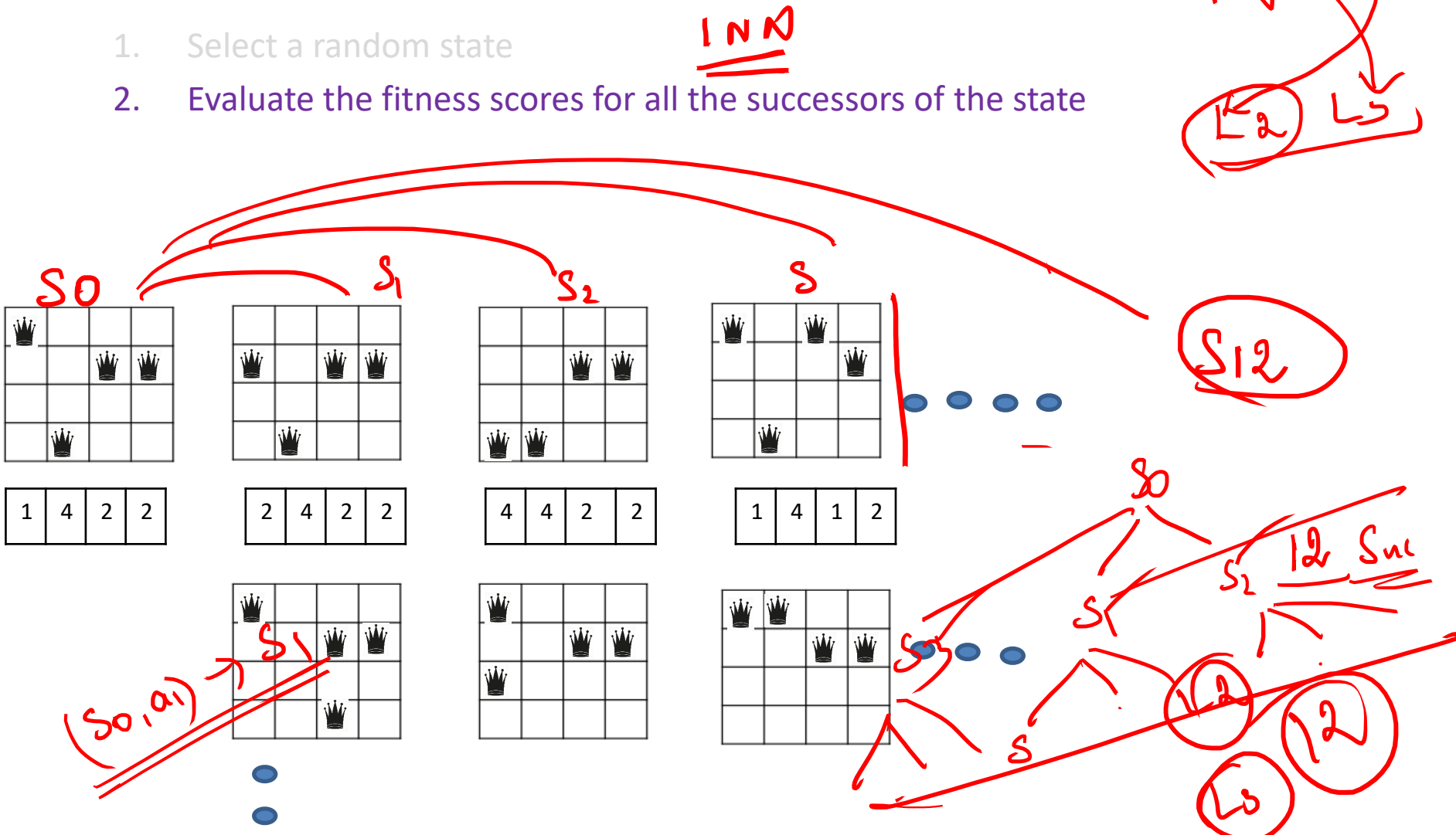
- Q1-Q2 - NC
- Q1-Q3 - NC
- Q1-Q4 - NC
- Q2-Q3 - NC
- Q2-Q4 - C
- Q3-Q4 - C



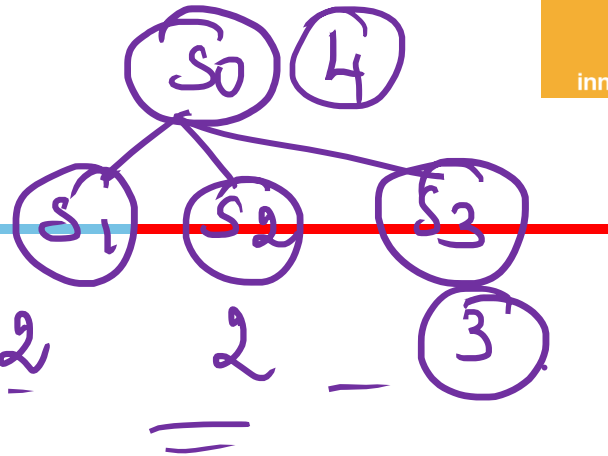
Note : Steps 3 & 4 in the above algorithm will be a part of variation of Hill climbing

Hill Climbing

1. Select a random state
2. Evaluate the fitness scores for all the successors of the state



Hill Climbing



S_1

4 \neq 2
4 \neq 2
4 \neq 3

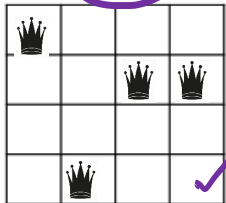
2

2

3

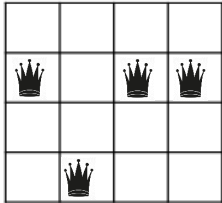
4 \neq 3

✓ S_0

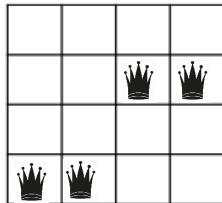


1 4 2 2 4

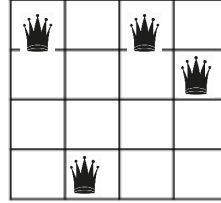
S_1



2 4 2 2 2



4 4 2 2 2



1 4 1 2 3

...

STOP

T C 1

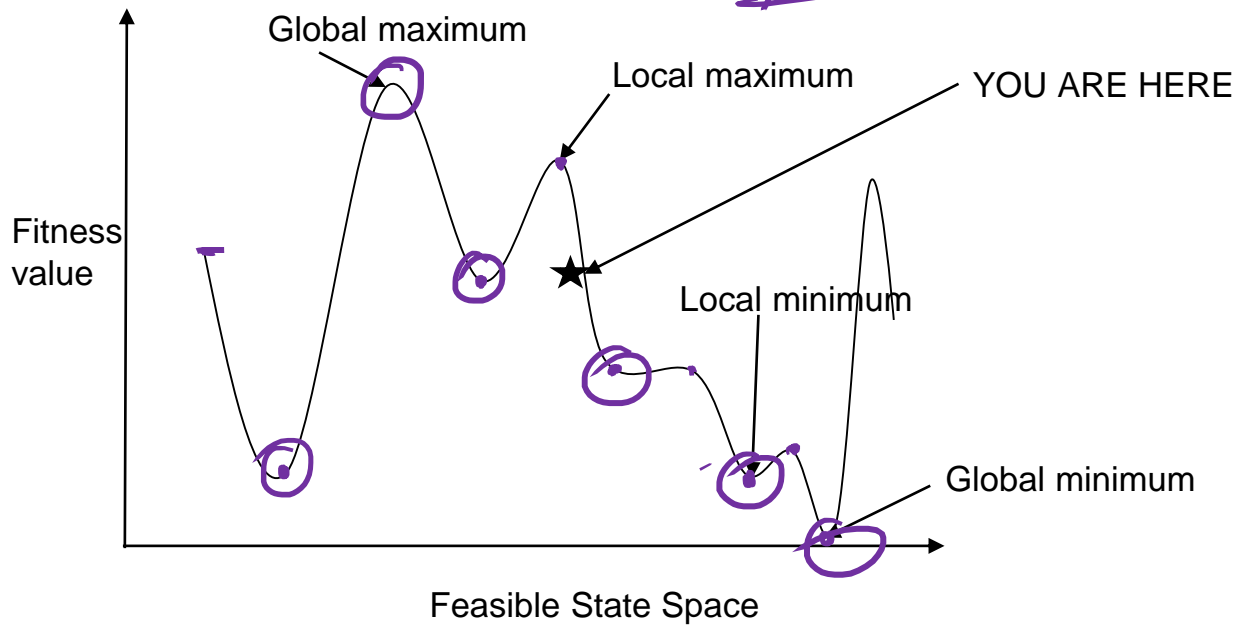
Local Maxima → Random Restart

Global maxing → 6
6 pairs NE

Hill Climbing



Countable R/RHA



Random Restart

- another*
1. Select a random state
 2. Evaluate the fitness scores for all the successors of the state
 3. Calculate the probability of selecting a successor based on fitness score
 4. Select the next state based on the highest probability
 5. Repeat from Step 2

Is

			♠
♠			
	♠	♠	

v1 v2 v3 v4

3	4	4	2	3
---	---	---	---	---

Ne

function HILL-CLIMBING(*problem*) **returns** a state that is a local maximum

current \leftarrow MAKE-NODE(*problem*.INITIAL-STATE)

loop do

neighbor \leftarrow a highest-valued successor of *current*

if *neighbor*.VALUE \leq *current*.VALUE **then return** *current*.STATE

current \leftarrow *neighbor*

Hill Climbing

Random Restart

1. Select a random state
2. Evaluate the fitness scores for all the successors of the state
3. Calculate the probability of selecting a successor based on fitness score
4. Select the next state based on the highest probability
5. Repeat from Step 2

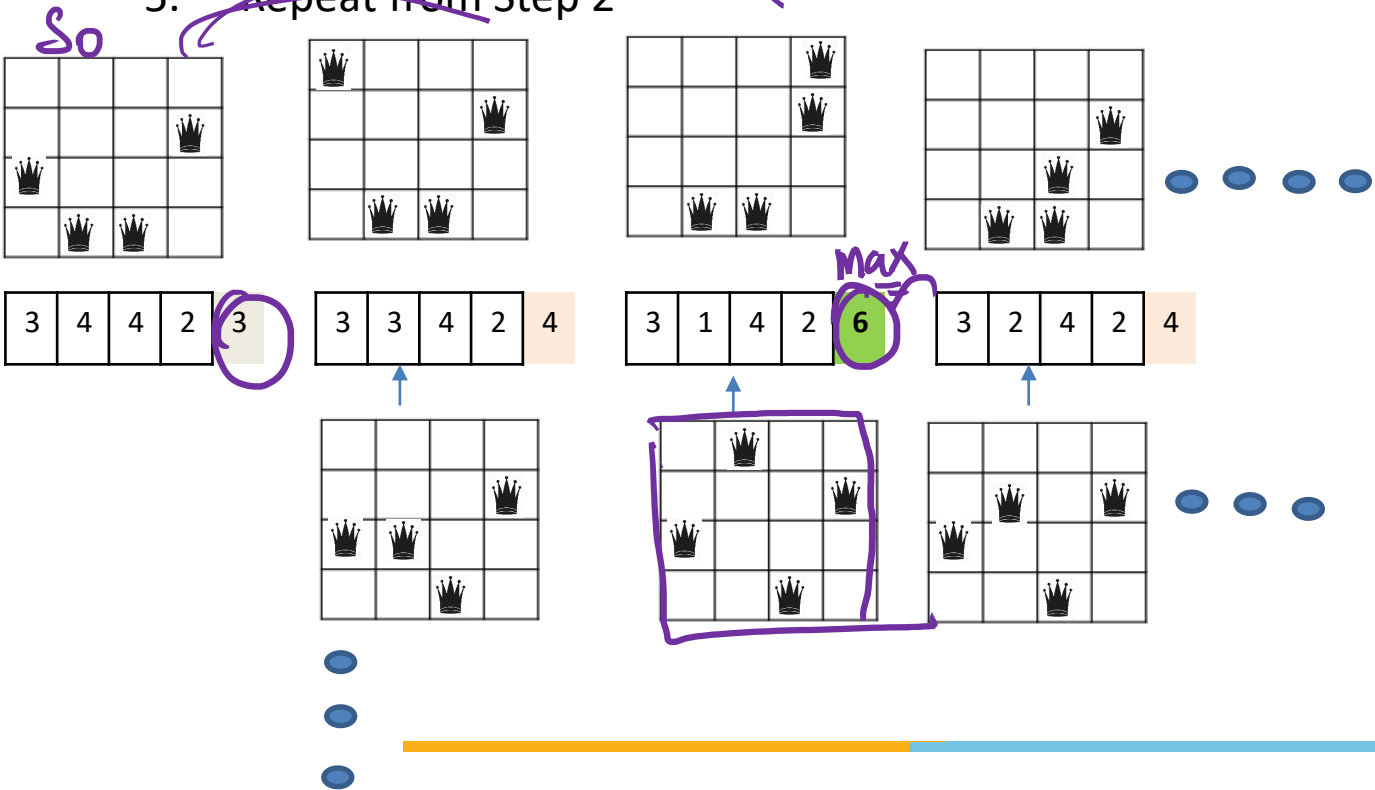
TC1

$$4 \neq 2$$
$$\leq 3$$

STOPS =

TC2

$$Gm/Gmb$$



TC2

12

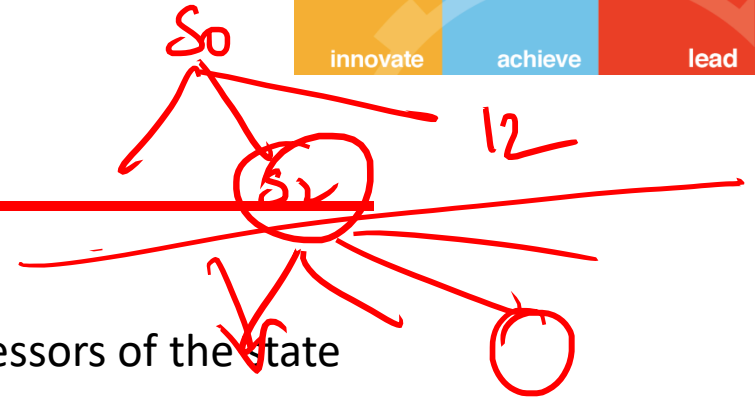
TC2

K=50

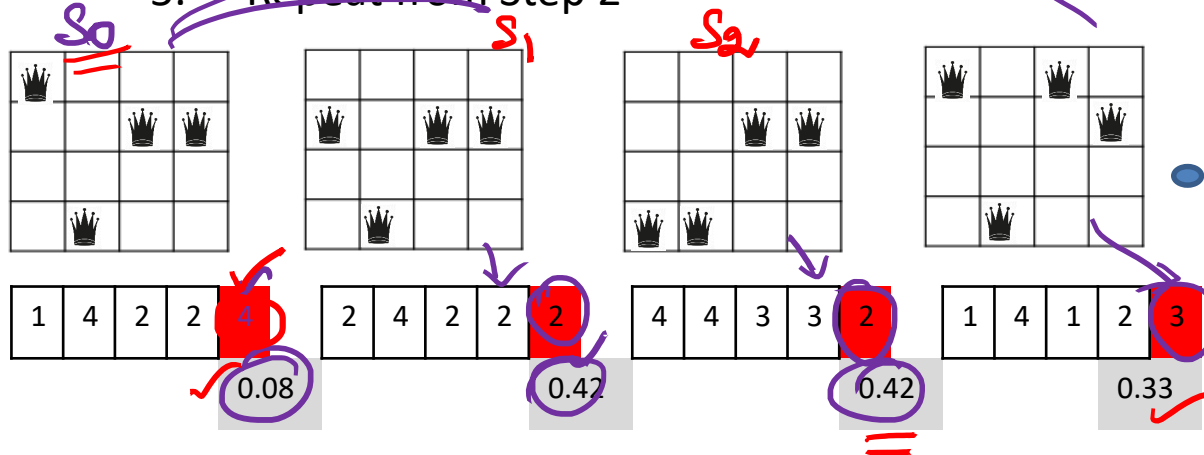
Stochastic Hill Climbing

next \leftarrow a randomly selected successor of *current*
 $\Delta E \leftarrow next.VALUE - current.VALUE$
if $\Delta E > 0$ **then** *current* \leftarrow *next*
else *current* \leftarrow *next* only with probability $e^{\Delta E/T}$

$f(n) - h(n)$ NC pairs Stochastic Hill Climbing →



1. ~~Select a random state~~
2. ~~Evaluate the fitness scores for all the successors of the state~~
3. ~~Calculate the probability of selecting a successor based on fitness score~~
4. Select the next state based on the highest probability
5. Repeat from Step 2



... 12

$f(n)$	$g(n)$	Prob	
1	2	2/12	0.16
2	5	5/12	0.42
3	4	4/12	0.33
4	1	1/12	0.08

$h(n)$ $\boxed{T} \rightarrow \text{Prob}$

12 N = {4, 2, 2, 3, 3, 2, 1, 3, 2, 1, 3, 2}

6 6

6 Question

4 to 5 Question

LOS
④
Gen

③ ACO

LOM
② Search

① (S)

→ PEAS

a) ST

b) $PC + h(n)$

Theory → TE 5 din + jnd

c) US, IDS, DFS, A*, IDA*

→ PSA con
is us, TM, PC
A

Any Alg
=

Local Beam Search

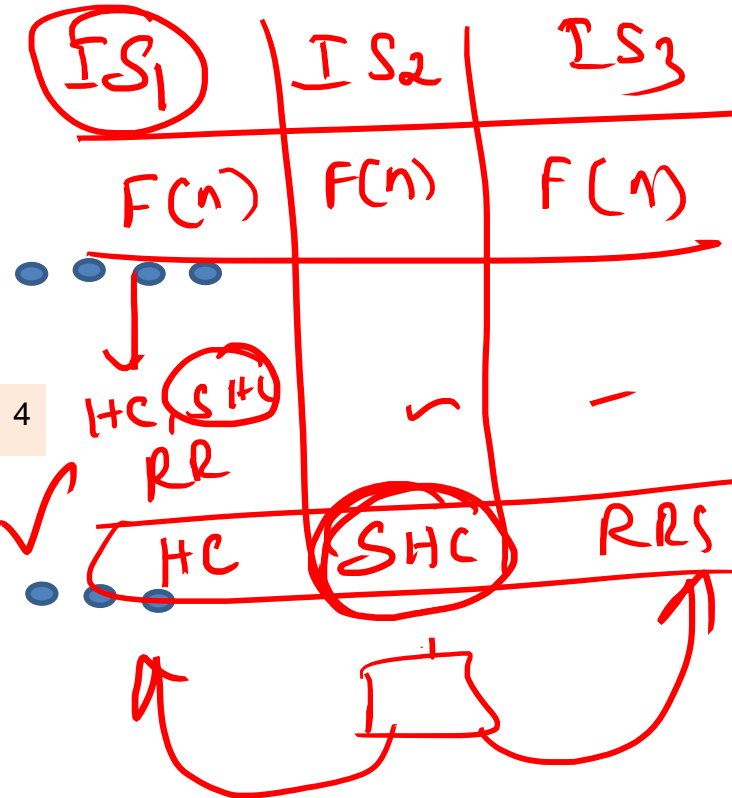
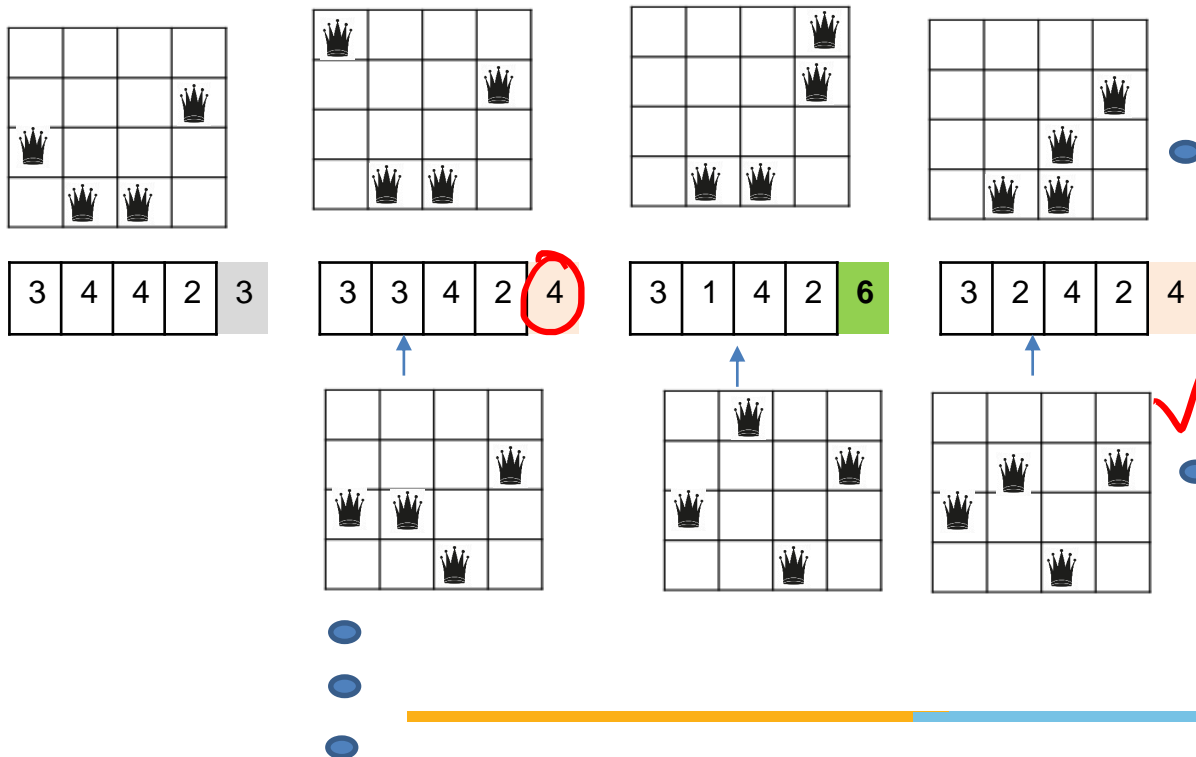
⑤ HD

Beam Search

One State



1. Initialize k random state
2. Evaluate the fitness scores for all the ~~successors~~ of the k states
3. Calculate the probability of selecting a successor based on fitness score
4. Select the next state based on the highest probability
5. If the goal is not found, Select the next 'k' states randomly based on the probability
6. Repeat from Step 2

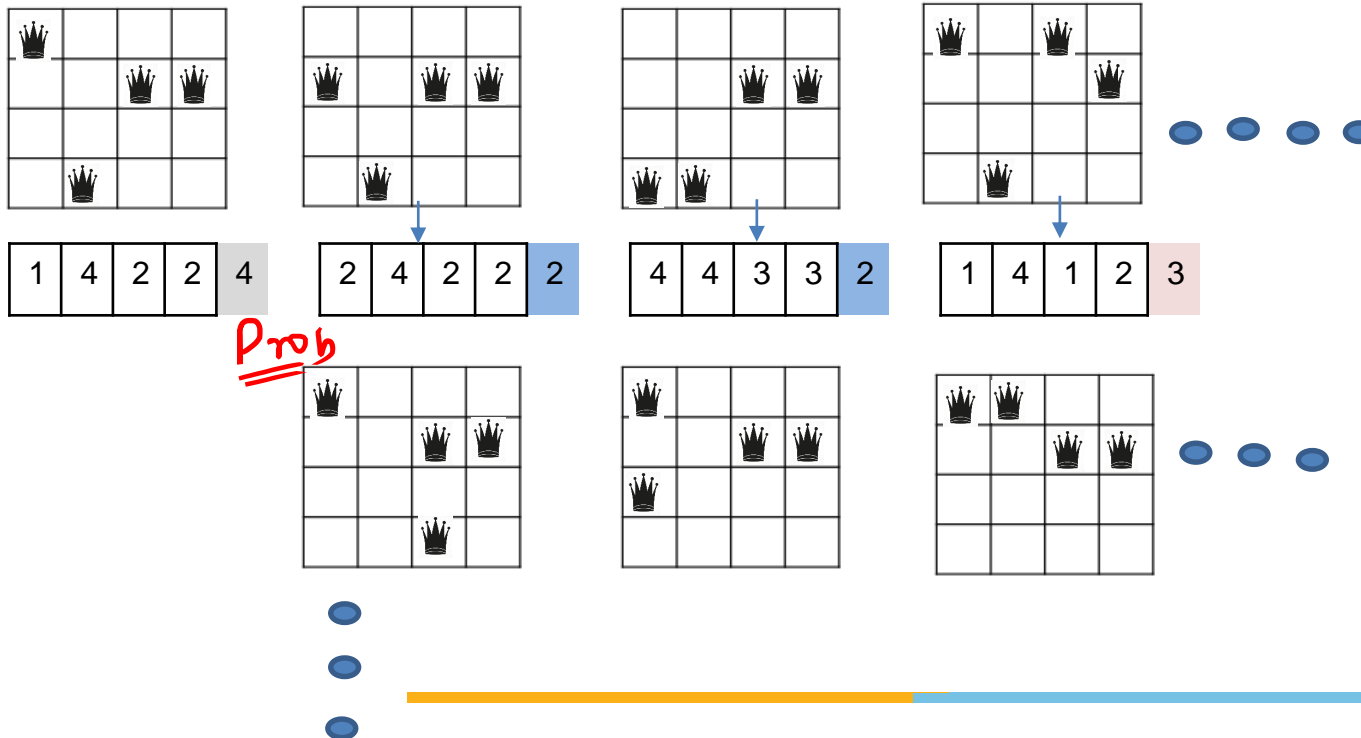


Stochastic Beam Search



Sample from 1st State

1. Initialize k random state
2. Evaluate the fitness scores for all the successors of the k states
3. Calculate the probability of selecting a successor based on fitness score
4. Select the next state based on the highest probability
5. If the goal is not found, Select the next 'k' states randomly based on the probability
6. Repeat from Step 2



Required Reading: AIMA - Chapter # 4.1, #4.2

Thank You for all your Attention

Note : Some of the slides are adopted from AIMA TB materials